

**THE EFFECT OF LOCAL SCHOOLS TRAFFIC ON CONGESTION
OF URBAN NETWORK**

BY

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A Thesis Presented to the
DEANSHIP OF GRADUATE STUDIES

KING FAHD UNIVERSITY OF PETROLEUM & MINERALS

DHAHRAN, SAUDI ARABIA

In Partial Fulfillment of the
Requirements for the Degree of

MASTER OF SCIENCE

In

CIVIL AND ENVIRONMENTAL ENGINEERING

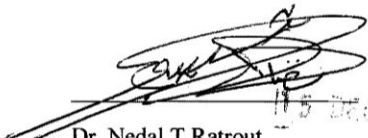
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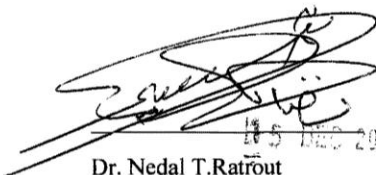
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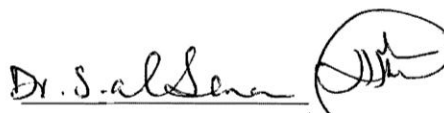
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DEANSHIP OF GRADUATE STUDIES


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2013

In the Name of Allah, Most Gracious, Most
Merciful

TO MY BELOVED FATHER,
MOTHER,
WIFE,
& AHMAD

ACKNOWLEDGMENTS

Acknowledgement is due to King Fahd University of Petroleum and Minerals for supporting this research.

I wish to express my deep appreciation to Prof. Nedal Ratrouf, who served as my major advisor, for his valuable advice and suggestions. Also, thanks are due to the other members of my thesis committee, Dr. Shukri Al-Senan and Dr. Khalaf Al-Ofi, for their useful suggestions.

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ABSTRACT

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OF URBAN NETWORK
Major Field : TRANSPORTATION ENGINEERING
Date of Degree : DECEMBER 2013

Schools in Saudi Arabia are of three levels: elementary, intermediary and secondary schools. It is thought that most of these schools built several years ago are based on the assumption that majority of its students will be walking to and from the schools since most of these schools were not provided with drop-off/pick-up areas. This research aimed to study the characteristics of trips attracted to local public boys' schools and suggest possible remedies to mitigate their adverse effect on the road network. It was found that the best variables which can be used to predict the number of attracted and parked vehicles are the number of students and number of employees. From the questionnaire analysis, it is obvious that 60% of overall students are suffering from congestion around schools but still 50% of them are coming to schools by passenger cars. Also, more than 60% of the students are ready to use the public transportation. With 33% of overall students ready to use the public transportation if it is free, 27% can pay an average of 50 SR.

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ملخص الرسالة

الاسم الكامل: خالد جمال عاصي

عنوان الرسالة: تأثير الحركة المرورية للمدارس على الازدحام المروري لشبكات الطرق في المدن الحضرية

التخصص: هندسة المواصلات

تاريخ الدرجة العلمية: ديسمبر, 2013

تقسم المدارس في المملكة الى ثلاثة مستويات و هي المدارس الابتدائية, المدارس المتوسطة و المدارس الثانوية. يعتقد ان معظم هذه المدارس التي بنيت من عدة سنوات تم بناؤها على افتراض ان غالبية طلابها سوف يعتمدون على المشي كوسيلة للتنقل من و الى المدرسة لأن معظم هذه المدارس لم تزود بمناطق نزول و ركوب الطلاب. هذا البحث يهدف الى دراسة خصائص الرحلات الى لمدارس البنين الحكومية في منطقة الظهران و الخبر واقتراح الحلول الممكنة للتخفيف من تأثيرها السلبي على شبكة الطرق. لقد وجد ان افضل المتغيرات التي يمكن استخدامها للتنبؤ بعدد السيارات المتوقفة و المجذوبة الى المدرسة هي عدد الطلاب و عدد الموظفين. تظهر نتائج تحليل الاستبيانات ان 60% من العدد الكلي للطلاب يعانون من الازدحام حول المدارس و لكن لا يزال 50% منهم يأتون الى المدارس بالمركبات. بالاضافة الى ذلك, 60% من عدد الطلاب الكلي مستعدون لاستخدام و سائل النقل العام بحيث 33% من عدد الطلاب الكلي مستعدون لاستخدام و سائل النقل العام اذا كانت من دون مقابل و 27% من عدد الطلاب الكلي على استعداد لدفع بما متوسطه 50 ريال سعودي كحد اعلى مقابل هذه الخدمة.

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CHAPTER 1

INTRODUCTION AND LITERATURE REVIEW

1.1 Introduction

The main adverse effects of congestion of urban networks are slower speeds, increased travel times and queuing. This affects negatively the economy and prosperity of any nation. Congestion also has adverse impacts on safety and quality of life. Even if traffic along the road network is acceptable at the present time, congestion is usually expected in the future due to natural growth of the population [1]. Major traffic generators such as malls, hospitals and schools contribute significantly to urban congestion. This research aimed to study and understand the trip attraction of public schools, and ultimately suggests some remedies to mitigate its contribution to urban congestion in the study area.

1.2 Problem Statement

Schools in Saudi Arabia are of three levels: elementary, intermediary and secondary schools. Most of these schools built several years ago are based on the assumption that the majority of its students will be walking to and from the schools since most of these schools were not provided with drop off/pick up areas. Consequently, the possible problem associated with vehicles congestion and parking demand was not taken seriously by the planners at that time. Unfortunately, nowadays one can easily observe that the majority of school trips are undertaken by vehicles in general and by passenger cars in

particular. It looks like that the major reasons for this shift to vehicular trips are the financial status of the population and the extremely poor public transit transportation system of the country. This research aimed to study the characteristics of trips attracted to local public schools and their effect on the surrounding road network, and ultimately suggests possible remedies to mitigate their adverse effect on the road network. This study focused on public schools since they constitute about 66% of the school population in the study area, and because their students usually share common socioeconomic characteristics. The study focused only on the AM peak at schools since it is the most critical peak period. The PM peak (at school closure) is not critical as the AM peak since a considerable number of schools and different classes within the same schools end at different times. Therefore, departure times at all schools are spread over long time periods to lessen the effect of congestion. This also makes manual data collection in the afternoon extremely hard.

1.3 Objectives

The specific objectives of this research were as follows:

1. To study the characteristics of the attracted trips to local public schools.
2. To investigate the socioeconomic factors of students that may affect the trip attraction to boys' public schools in the study area.
3. To develop trip and parking generation models to understand the traffic impact of boys' public schools that can be used locally to predict the vehicle trips attracted to public schools.
4. To test the developed model on samples of public schools in the local area.

5. To review possible methodologies and techniques which are promising in mitigating the problem of congestion around schools.
6. To assess the acceptance of some of these methodologies by students and their parents by conducting a questionnaire survey.

1.4 Literature Review

Increase in demand for transportation plays a leading role in creating congestion in urban areas. Economic growth and social development increase mobility in cities and promote the use of private vehicles. People need or desire access to numerous activities in more complex urban environments, therefore, an increase in demand for transportation is inevitable. However, transportation infrastructure is limited and is not always used in an optimal fashion. There are number of factors influencing demand for travel and result in congestion in the short or long run, such as [1]:

- Socioeconomic growth.
- Increase in urban population.
- Car ownership and dependency.
- Land uses.
- Travel patterns.
- Public transport operations.
- Urban freight transport and goods delivery.
- Parking.

Saudi Arabia has experienced a rapid economic growth since the oil boom, resulting in a dramatic increase in the number of registered vehicles from 144,768 in 1970 to 12,895,705 in 2010 according to the available most recent data [2]. This growth in motorization was accompanied with an increase in the size of the road network in the Kingdom. As a result of this remarkable growth in motorization and the road network size, traffic accidents and congestion have become a serious problem facing this country. The number of traffic accidents in 2011 was 544,179 accidents with a total number of 7153 fatalities according to the available most recent data [3]. Trip generation rates from elementary, middle and high schools have all increased noticeably over the past years because of several factors such as reduced walking and biking as well as reduced school bus programs. The Trip Generation report, 8th Edition of 2008 [4] estimates that elementary schools generate about 1.29 trip ends per student per weekday, with 0.45 during the AM peak hour and 0.28 during the afternoon peak hour. This is a significant increase from the rate of 1.02 trip ends per student per weekday, with 0.30 during the AM peak hour and 0.26 during the afternoon peak hour in Trip Generation report, 6th Edition, 1997 [5]. The main independent variables in ITE (Institute of Transportation Engineers) are the number of students, the number of employees and gross floor area.

According to the Trip Generation report, 8th Edition [4], the best fit regression curves are accepted for any land use only when each of the following three conditions is met:

- 1- The R^2 is greater than or equal to 0.50.
- 2- The sample size is greater than or equal to 4.
- 3- The number of trips increases as the size of the independent variable increases [4].

Trip generation manual for Riyadh city provides the trip rates for different land uses in Riyadh. The minimum sample size used to predict the trip rates for any level of public schools is 21 schools.

The trip rates for morning peak hour of each level of schools and independent variables for different manuals used are shown in Tables 1.1 to 1.8 [6].

Table 1.1: Riyadh trip rates for secondary public schools (boys and girls) (morning peak)

Independent variable	Sample size	Average rate	Regression model	R ²
Total Floor Area (m ²)	24	0.116	y= 0.101x+47.5	0.50
Number of Students		0.654	y=0.7609x-60.7	0.58

Table 1.2: Riyadh trip rates for intermediary public schools (boys and girls) (morning peak)

Independent variable	Sample size	Average rate	Regression model	R ²
Number of Students	31	0.75	y=0.5884x+78.1	0.68

Table 1.3: Riyadh trip rates for primary public schools (boys and girls) (morning peak)

Independent variable	Sample size	Average rate	Regression model	R ²
Total Floor Area (m ²)	22	0.117	y= 0.056x+145.4	0.50
Number of Students		0.66	y=0.748x-45.3	---
Number of Teachers		9.96	y=6.227x+126.7	0.79

Source: The Ministry of Municipal and Rural Affairs, Saudi Arabia, Trip Generation Manual for Riyadh City, 2009 [6].

Table 1.4: ITE trip rates for secondary schools (morning peak)

Independent variable	Number of studies	Average rate	Regression model	R²
Number of Students	68	0.42	Not given	Not given
Number of Employees	53	4.68	Not given	Not given
1000 sq. ft. Gross Floor Area	44	3.06	Not given	Not given

Table 1.5: ITE trip rates for intermediary schools (morning peak)

Independent variable	Number of studies	Average rate	Regression model	R²
Number of Students	25	0.54	Not given	Not given
Number of Employees	21	5.30	$y=9.25x-300.80$	0.54
1000 sq. ft. Gross Floor Area	21	4.35	Not given	Not given

Table 1.6: ITE trip rates for primary schools (morning peak)

Independent variable	Number of studies	Average rate	Regression model	R²
Number of Students	48	0.45	$\ln(Y) = 1.14\ln(X) - 1.86$	0.50
Number of Employees	50	5.37	$Y = 7.91x - 127.63$	0.66
1000 sq. ft. Gross Floor Area	58	5.20	$\ln(Y) = 1.20\ln(X) + 0.66$	0.63

Source: Institute of Transportation Engineers, Trip Generation, 8th edition, 2008 [4].

Table 1.7: Dubai trip rates for secondary schools (morning peak)

Independent variable	Number of studies	Average rate	Regression model	R²
Number of Students	Not given	0.52	Not given	Not given
100 sq. m. Gross Floor Area	Not given	1.086	Not given	Not given

Table 1.8: Dubai trip rates for primary and intermediary schools (morning peak)

Independent variable	Number of studies	Average rate	Regression model	R²
Number of Students	Not given	0.43	Not given	Not given
100 sq. m. Gross Floor Area	Not given	2.750	Not given	Not given

Source: Roads and Transport Authority, G.o.D., Dubai Trip Generation and Parking Rates 2013 [7].

One of the important types of recurring congestion all over the world is the school-related traffic congestion which has adverse effects on the safety of the students, parents, teachers, motorists and residents [8]. There are many studies about congestion around schools, factors that affect the modal choice of students, and possible solutions to reduce the impact of congestion around schools.

One study which was done at 34 California public elementary schools showed that walking and biking rates in the schools are influenced negatively by school size and positively by neighborhood population density as shown in Figure 1.1 below [9-10].

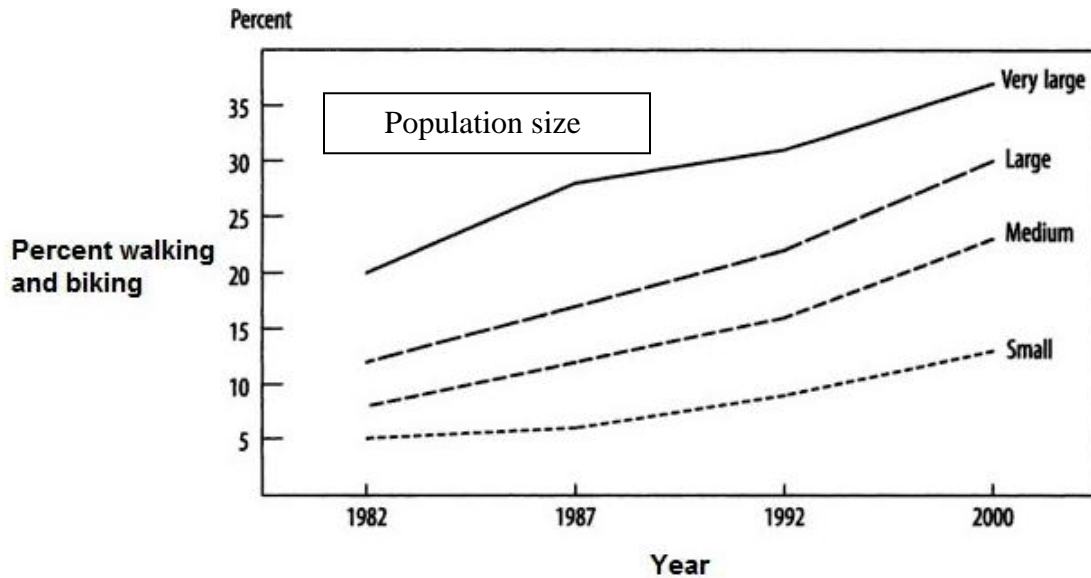


Figure 1.1: Percent of daily travel in congestion, by regional size group [10]

Another study which was done by Black et al. [11] found a significant relationship between mode choice and the distance from school to home. They concluded that the probability of traveling by automobile instead of by foot increased from 20% at 0.5 mi distance to 50% at 1.25 mi and 80% at 2 mi.

Cynecki and Brownlee [12] related the reduction in walking and increased congestion around schools to the number of factors such as:

- Increased size of the school: larger population of schools results in larger attendance boundaries with longer walking distances.
- School placement within the attendance boundary: schools must be in the center of the attendance boundary to minimize walking distances.
- Traffic circulation and connectivity: schools should have vehicle access from at least two different streets since more access will result in less congestion and

more efficient traffic operation. The access of pedestrians and bicycles should be provided from all points around the schools.

- Lack of sidewalks: the presence of safe sidewalks around the school will encourage the parents to allow their children to walk to the school.
- Pick-up and drop-off areas for school buses and parents.
- Inadequate parking.
- Teenage drivers at high schools, who are the least experienced drivers, are more prone to crashes.

Another study concluded that the walk travel time is the most significant factor affecting the decision to walk to the school [13].

Slipp and Hummer [14] made a study to determine the trip generation rates for public high schools in North Carolina and compare them with those presented in ITE trip generation report, using students, employees and gross floor area as independent variables. They found that the rates are higher than those of the ITE trip generation. Dean et al. [15] conducted a study to determine the demand for pick-up/drop-off facilities to develop retrofit strategies for existing schools in the city of Surrey in British Columbia and design guidelines for new development. The data collection of the project contained a combination of school site surveys and questionnaires for twenty elementary schools. The number of vehicles arriving and departing was recorded each minute over a period of one hour. The purpose of the questionnaire was to determine the number of vehicles picking up or dropping off by asking the students about the mode they used in coming to the school. Various trend types (linear, logarithmic, power and exponential) were checked

using the number of students as an independent variable. It was concluded that the best model is the power trend type since it has the highest R^2 .

Transportation management strategies generally exclude extensive infrastructure investments aimed at expanding roadway capacity. Instead, these strategies focus on:

1. Management of travel demand to reduce the severity and duration of circumstances where travel demand exceeds existing roadway capacity. Modifications to travel demand can include adjustments to travel time (by time-of-day and/or day-of-week), travel route, trip distance (through changes in trip origins and destinations), and vehicle occupancy.
2. Management of existing corridor capacity to address locations where relatively minor improvements to the roadway network or highway operations will help address temporary or long-term capacity bottlenecks. Temporary bottlenecks include those caused by incidents, weather, and construction factors.

The congestion management system (CMS) is intended to provide information on transportation system performance and identify alternative actions to alleviate congested roadway conditions. Figure 1.2 presents the steps for CMS [16].

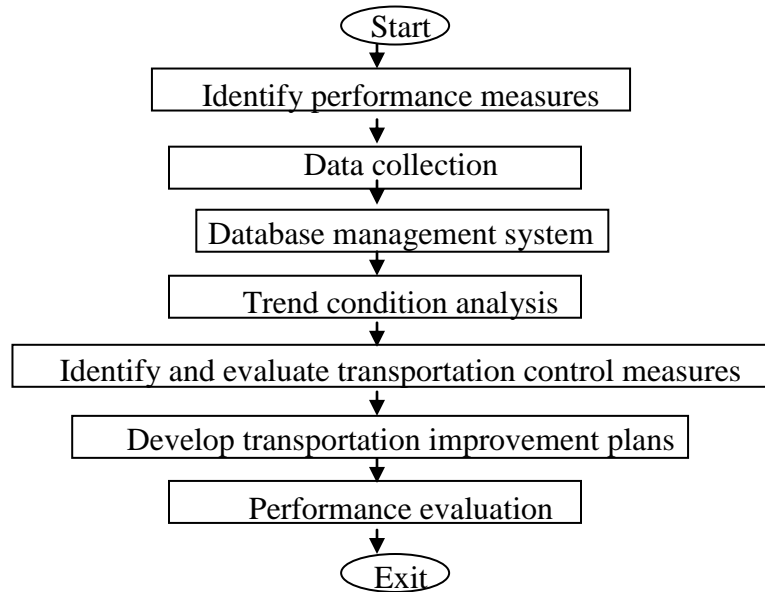


Figure 1.2: Possible CMS flowchart [16]

There are three types of measures which are used to reduce congestion in cities:

1- Short-term measures such as:

- Change of traffic signal regulation systems.
- The systems for automatic traffic control (such as changeable message signs).

2- Medium-term measures which include major traffic and technical works that significantly improve the current conditions.

3- Long-term measures that permanently change the current conditions such as construction of new streets [17].

The common reasons which make the parents take their children to school by car are traffic hazards, time constraints, bad weather, and strange danger [18].

The most efficient solution to the problem of traffic congestion around schools is encouraging walking and biking to school by enhancing safety around schools. There are number of factors that affect the overall safety around schools, such as physical setting, school population characteristics, and law enforcement [19].

One of the studies done by Bradshaw [20] in the north-western suburbs of Leeds suggested ways to make walking and cycling more attractive, which can be summarized as follows:

- Provide widened pavement in order to redress the problems caused by children or parents who sometimes have to walk on the road.
- Lessen the numbers of cars within close proximity of the schools by providing traffic free zones around schools.
- Provision of drop-off and collection points for those walking and/or catching buses at which larger groups of children could gather and be walked to the school by a responsible adult.
- Make parents realize that in driving their children to the school, they are contributing to the problems of congestion and road safety.
- Provide cycle lanes to the schools.

According to the Astralian Department of Transport [21], there are several attributes that are extremely important in providing safety around schools, such as:

- Traffic speeds should be low, desirably 40 km/h or less.

- Providing paths on the school side of the road for children walking and cycling.
- Sight lines for drivers to see children and to be seen by children should be clear at intersections and all places where children might cross a road.
- Road crossing places for children should be safely located and adequately signed.
- All pedestrian and bicycle access ways should be free from visibility constraints.

One of the suggested solutions to reduce congestion around schools and encourage walking/biking to school is to enhance safety by implementing safe routes to school (SRTS) programs. Safe Routes to School is a national and international movement to create safe, convenient, and fun opportunities for children to bike and walk to and from the schools. The effectiveness of this technique was measured by Zhao and Zhou [22] with the data collected before and after the implementation of the SRTS programs in Hillsborough County, Florida. He indicated that walking/biking rates improved significantly.

La Vigne [8] suggested specific responses to reduce traffic congestion around schools, such as:

- *Educating parents:* educate parents about their children using alternative transportation modes to and from the school, the dangers and legal consequences of traffic violations as well as the role that parents can play in reducing congestion and increasing student safety by following the rules of the road.
- *Encouraging carpooling.*
- *Mapping out safe pedestrian routes.*

- *Instituting school busing:* this is an effective means of reducing the number of children taken to schools by car.
- *Altering drop-off and pick-up rules:* altering the time during which parents can drop off or pick up their children or by staggering bell times.
- *Strategically funneling traffic:* this is done by directing different types of traffic to different locations around the school property. For instance, locate staff parking in another area for pick-up and drop-off.
- *Establishing parking zones.*
- *Synchronizing traffic signals.*

To enhance the student loading (pick-up and drop-off) efficiency and safety, Qualls [23] suggested to use the automatic barcode reading technology in which the entrance of the school driveway is equipped with a barcode reader that reads a barcode decal on each pick-up vehicle entering the driveway. Then, the students to be picked up are staged in the classroom in succession according to the scans of the barcode decals. Another ITS (Intelligent Transportation Systems) technique suggested by Sarasua et al. [24] is to enhance after-school traffic operation at an elementary school in Clemson, South Carolina. The technology used is the radio frequency identification (RFID) system which provides advanced notification of the approaching vehicles and their intentions so that students could be loaded into cars more efficiently. It was concluded that the usage of this new technology will increase the school's capacity to handle after-school pick-ups by 14%, which results in decreasing the vehicle wait times, shorter queue lengths and more efficient use of passenger pick-up area. Also, one of the most widely used components of ITS is the Advanced Traveler Information System (ATIS). It includes Variable Message

Signs (VMS), route guidance, telephone information, and radio systems. These techniques assist motorists in making more informed decisions to avoid congestion by their pre-route and en-route path selection.

Parking supply and demand management strategies provide efficient solutions to many of the problems facing the transportation system, such as delay reduction, capacity utilization and travel time reliability. Many studies and manuals were conducted and developed to predict the parking demand for different land uses [25]. ITE Parking Generation report provides the parking demand rates for the three levels of schools for different independent variables such as number of students and the number of employees as shown in Table 1.9.

Table 1.9: Parking demand rates for schools [26]

Level of school	Sample size	Independent variable	Average peak period parking demand
Elementary	6	Students	0.28 veh/student
Middle	Not given	Students	0.11 veh/student
		Employees	1.2 veh/employee
High	3	Students	0.09 veh/student

Source: McCourt, Institute of Transportation Engineers, Parking Generation, 3rd Edition, 2004 [26].

After applying a specific solution to the problem of congestion around schools, some of the measures that can be used to evaluate the effectiveness of this solution are as follows [8]:

- Fewer vehicles around schools.

- Reduced time spent by the parents in dropping off and picking up their children.
- Fewer vehicular crashes around schools.
- Fewer pedestrian injuries and deaths around schools.
- Fewer traffic violations around schools.
- Lower percentage of parents using cars to take their children to the school.
- Improved perceptions of congestion among parents and staff.

In summary, there are different reasons for the congestion around schools, such as school placement, traffic circulation, lack of sidewalks, pick-up and drop-off areas and inadequate parking. From the literature, there are a wide range of remedies that can be used to reduce congestion around schools, such as:

- Educating parents
- Encouraging carpooling
- Instituting school busing
- Altering drop-off and pick-up times
- Establishing parking zones
- Synchronization of traffic signals

1.5 Study Area

This research was conducted on samples of public schools located in Al-Khobar and Dhahran cities in Saudi Arabia. Convenience is the main reason for selecting this area as a case study. The total population of Al-Khobar and Dhahran cities is 587,965 capita [27], which represents nearly 2.2% of the total population of Saudi Arabia. The total

number of all levels of public, private and international schools in the study area are presented in Table 1.10 [27].

Table 1.10: School population in the study area [27]

School type	Total number
Public Secondary	18
Public Intermediary	25
Public Primary	45
All Private	45
Total	133

Source: Census Report for 2010 , Saudi Arabia, 2010 [27].

This study focused on public schools since they constitute about 66% of the total school population in the study area, and because their students usually share common socioeconomic characteristics.

1.6 Data Collection

Because of the lack of financial resources and human manpower, data were collected at only 30 different schools in Al-Khobar and Dhahran cities which represent 34% of public schools via manual traffic counts. All of the schools were selected randomly. All of the businesses conducted at these particular schools occur during the morning peak hour period. Therefore, the trip generation data collection and analysis was limited to the AM peak hour period only. The study focused only on the AM peak at schools since it is the most critical peak period. The PM peak (at school closure) is not critical as the AM peak since a considerable number of schools and different classes within the same school end at different times. Therefore, departure times at all schools are spread over long time

periods to lessen the effect of congestion. This also makes manual data collection in the afternoon extremely hard.

All of the data were collected during March and April of 2013 on typical weekdays which are Saturday, Sunday, Monday, Tuesday, Wednesday and Thursday. The number of drop-off and parked vehicles was documented at each of these schools between the hours of 6:15 AM to 7:15 AM, which represents the AM peak hour, since the schools start at 7:00 AM. Figure 1.3 indicates the approximate locations of schools included in this analysis.



Figure 1.3: Approximate schools locations

Data for the following variables were considered at all 30 schools included in this study:

- Number of students for each school
- Number of employees for each school
- Number of classes
- Total floor area (m^2): the sum (in square meter) of the built area of each floor level.
- Total lot area (m^2): defined as the area (in square meter) of the built and unbuilt land owned by the school.
- Airline distance to nearest school (m)
- Airline distance to nearest arterial (m)
- Airline distance to nearest collector (m)
- Airline distance to central business district (m): The CBD or Central Business District is the focal point of a city. It is the commercial, office, retail, and cultural center of the city and usually is the center point for transportation networks. According to the municipality of Al-Khobar, the CBD was presented as a black circle in Figure 1.3. It is hypothesized that as the value of this variable increases, the population density decreases, and consequently, the number of schools are less and the distance between schools will be large, which necessitates vehicular trips rather than walking to school.

The number of students, employees and classes were determined by asking the headmaster directly. The total lot area which represents the built and unbuilt land owned by the school, and the total floor area which represents the area of all buildings of the

school, were both determined using the Google Earth software. Also, Google Earth was used to measure all the distances.

The collected data for each level of schools are summarized in a tabulated format as shown in Tables 1.11 to 1.13 below.

Table 1.11: Collected data for secondary schools

Reference number	School	Drop-off vehicles	Parked vehicles	Total Attracted vehicles	Number of Students	Number of Employees	Number of Classrooms	Lot Area (m ²) (1000 sq m)	Floor Area (m ²) (1000 sq m)	Distance to nearest school (m)	Distance to nearest arterial (m)	Distance to nearest collector (m)	Distance to CBD (m)
1	Al-Kisa'i	68	131	290	864	51	23	10.36	5.40	1364	25	40	2614
2	Al-Thuqba	89	70	240	721	45	21	14.25	5.70	1335	1115	44	1828
3	Ibn Senaa'	191	72	250	537	55	19	11.50	8.4	1667	100	50	1651
4	Al-Oroba	84	73	157	310	24	9	13.30	5.84	4803	510	435	8200
5	Al-Khobar	150	109	235	500	53	13	19.364	6.80	130	440	65	741
6	Al Ya'qobi	70	51	121	293	37	12	10	3.227	130	440	65	830
7	Ibn Al-Qassem	68	39	107	191	30	6	6.43	3.638	1226	210	150	1138
8	Jabal Al-Noor	144	114	245	448	48	10	9.25	5.775	3149	910	294	3512
9	Al-Khozama	88	84	172	251	30	9	5.265	4.80	4201	574	412	7829
10	Tayba	61	70	149	414	32	13	7.60	3.23	120	20	40	2336
	Average	101	81	196.6	452.9	13.5	40.5	10.73	5.28	1812	434	160	3068
	Range	61-191	39-131	107-290	191-864	24-55	6-23	6.43-19.364	3.22-8.4	120-4803	20-1115	40-435	741-8200
	Coefficient of variation	0.44	0.35	0.32	0.47	0.28	0.42	0.39	0.31	0.88	0.80	0.95	0.85

*Arterial is any divided two-way road with at least two lanes in each direction or higher classification.

Drop-off vehicles: the number of vehicles which came to drop the students and left.

Parked vehicles: is the number of vehicles which parked near the school for school purposes.

Total attracted vehicles: is the summation of drop-off vehicles and parked vehicles.

Table 1.12: Collected data for intermediary schools

Reference number	School	Drop-off vehicles	Parked vehicles	Total Attracted vehicles	Number of Students	Number of Employees	Number of Classrooms	Lot Area (m ²) (1000 sq m)	Floor Area (m ²) (1000 sq m)	Distance to nearest school (m)	Distance to nearest arterial (m)	Distance to nearest collector (m)	Distance to CBD (m)
11	Al-Khobar	285	48	333	512	52	15	19	5.68	900	305	125	1910
12	Ibn-Muthaffar	279	42	321	590	48	15	15.328	4.90	2226	750	350	8607
13	Al-Farabi	226	35	261	422	46	15	14.042	5.33	976	337	157	2330
14	Ashbielia	168	42	270	470	40	23	5.49	4.44	903	340	30	1667
15	Ibn Al-Awwam	157	38	240	417	42	15	13	4.20	1530	213	44	874
16	Eskan Al-Khobar	58	38	150	375	40	14	7.15	3.576	562	528	50	2543
17	Al-Qadessiyya	39	18	57	200	18	10	2.295	0.765	1507	116	45	2317
18	Al-Jesser	130	30	160	280	21	7	5.265	1.80	905	610	412	7829
19	Al-Thuqba	171	28	199	389	37	10	11.10	4.10	1120	600	110	1176
20	Al-Dhahran	195	28	223	375	36	12	11.75	3.30	564	1150	125	5863
	Average	171	35	221	403	14	38	10.44	3.809	1119	495	145	3511
	Range	39-285	18-48	57-333	200-590	21-52	7-23	2.295-15.328	0.765-5.68	562-2226	213-1150	30-350	874-8607
	Coefficient of variation	0.48	0.25	0.38	0.27	0.29	0.32	0.50	0.40	0.43	0.58	0.87	0.77

*Arterial is any divided two-way road with at least two lanes in each direction or higher classification.

Drop-off vehicles: the number of vehicles which came to drop the students and left.

Parked vehicles: is the number of vehicles which parked near the school for school purposes.

Total attracted vehicles: is the summation of drop-off vehicles and parked vehicles.

Table 1.13: Collected data for primary schools

Reference number	School	Drop-off vehicles	Parked vehicles	Total Attracted vehicles	Number of Students	Number of Employees	Number of Classrooms	Lot Area (m ²) (1000 sq m)	Floor Area (m ²) (1000 sq m)	Distance to nearest school (m)	Distance to nearest arterial (m)	Distance to nearest collector (m)	Distance to CBD (m)
21	Ubai Ibn Thabet	229	40	260	410	43	20	18.594	4.18	1323	722	206	8702
22	Al-Abbas	111	44	170	430	45	21	8.231	3.30	532	145	45	1803
23	Al-Seddeq	205	30	220	375	31	15	11.66	3.141	925	236	70	1762
24	Ubai Ibn Kaab	129	27	156	311	28	14	2.299	1.584	507	328	30	1003
25	Hitteen	109	45	250	658	48	26	6.636	3.661	87	223	42	2565
26	Yohai ibn Mo'en	84	16	100	203	17	8	0.284	0.284	3441	272	694	8524
27	Hajr	50	10	60	115	12	6	2.58	1.485	1592	250	130	3357
28	Mohammadia	136	29	165	440	32	18	3.769	3.136	443	30	31	1098
29	Muath Ibn Jabal	93	20	113	188	20	7	5.106	1.592	507	435	35	1330
30	Dhahran	188	35	223	420	37	18	10.144	4.042	575	1310	747	5740
	Average	133	30	172	355	15	31	6.93	2.64	993	395	203	3588
	Range	50-229	10-45	60-260	115-658	12-48	6-26	0.284-18.594	0.284-4.18	87-3441	30-1310	30-747	1003-8524
	Coefficient of variation	0.43	0.40	0.39	0.44	0.39	0.43	0.79	0.50	0.93	0.89	1.30	0.80

*Arterial is any divided two-way road with at least two lanes in each direction or higher classification.

Drop-off vehicles: the number of vehicles which came to drop the students and left.

Parked vehicles: is the number of vehicles which parked near the school for school purposes.

Total attracted vehicles: is the summation of drop-off vehicles and parked vehicles.

1.7 Methodology

The ultimate objective was to evaluate the traffic impact of schools, and this was accomplished in three parts. The first part was modeling of trip and parking generation to estimate the traffic generated in the school, which can help in avoiding the problem of congestion around the proposed schools by providing them with adequate facilities, such as pickup/drop-off areas, parking lots and wider surrounded roads. On the other hand, it can help in predicting the number of attracted vehicles to the existing schools using the rates only rather than counting the number of vehicles. This was conducted by observing the number of vehicular trips and relating it to the number of factors, such as number of students, number of employees, number of classrooms and floor area. The second part was assessing the acceptance by students and their parents of some of the remedies for mitigating congestion around the existing schools. These remedies were taken from extensive literature review about the problem of congestion around schools and the suggested remedies to solve this problem. Some of these remedies which were suggested in the literature are as follows:

- Instituting school busing
- Providing parking lots
- Segregation times for the start of the schools
- Distance learning
- Encourage students to walk by providing safe routes to the school

This part was mainly conducted via questionnaire distributed to the students and their parents to assess the acceptance of such remedies as seen by the students and their

parents. The final stage was an engineering exercise to show and prove that a simple straightforward engineering solution, such as optimizing signal timing plan at intersections near schools, can reduce congestion considerably. The detailed methodology for conducting these parts is as follows:

1. Identify and determine the public school population in the study area.
2. Conduct a literature review of all possible remedies to mitigate the adverse effect of schools on congestion of urban networks.
3. Select a representative sample.
4. Study the trip attraction characteristics through manual counting of vehicles attracted to schools.
5. Conduct a survey questionnaire on students to study the socioeconomic characteristics of students which might affect the trip making characteristics.
6. Study the traffic peaking characteristics of a sample of major streets surrounding some of the studied schools through automatic traffic counts.
7. Conduct a questionnaire survey with students' parents to evaluate the appropriateness of some remedies in alleviating school congestion in the study area.
8. Draw up conclusions and recommendations.

CHAPTER 2

MODELING OF VEHICULAR TRIPS

ATTRACTION

2.1 Data Analysis and Discussion

Regression analysis is a statistical tool for the investigation of relationships between variables. Usually, the investigator seeks to ascertain the causal effect of one variable upon another.

In this study, single and multiple regressions were used to develop a model which predicts the number of attracted vehicles to schools using many of the dependent variables.

2.1.1 Experimental Design

A set of explanatory factors summarizing the characteristics of schools was evaluated. The relationship between these explanatory variables and the number of attracted vehicular trips (dependent variable) was thoroughly investigated through regression. The explanatory variables are of two types:

1. Variables which describe the characteristics of the school:
 - The number of students
 - The number of employees

- The number of classrooms
- Total floor area: the sum (in square meter) of the area of each floor level
- Total lot area: defined as the area (in square meter) of the built and unbuilt land owned by the school

2. Variables which describe the accessibility to school:

- Airline distance to the nearest school
- Airline distance to the nearest collector
- Airline distance to the nearest arterial
- Airline distance to central business district (CBD)

For each dependent variable, the analysis was done four times for each level of schools as described below:

1. Data analysis for seven schools with intercept
2. Data analysis for seven schools without intercept
3. Data analysis for ten schools with intercept
4. Data analysis for ten schools without intercept

The purpose of analyzing seven schools out of ten is to use the remaining three schools in the validation process which can help in determining the best fit models. For each level of schools, the models which resulted from analyzing seven schools were compared with those that resulted from analyzing ten schools.

The same procedure was done when all schools were pooled together as described below:

1. Data analysis for 21 schools with intercept
2. Data analysis for 21 schools with intercept using two dummy variables to distinguish between the three levels of schools
3. Data analysis for 21 schools without intercept
4. Data analysis for 21 schools without intercept using two dummy variables to distinguish between the three levels of schools
5. Data analysis for 30 schools with intercept
6. Data analysis for 30 schools with intercept using two dummy variables to distinguish between the three levels of schools
7. Data analysis for 30 schools without intercept
8. Data analysis for 30 schools without intercept using two dummy variables to distinguish between the three levels of schools.

The data were analyzed using the SPSS software.

2.1.2 Data Analysis for Secondary Schools

2.1.2.1 Summary of Results

The correlation between the dependent variable (total attracted vehicles) and independent variables for secondary schools was introduced by the following matrix (Table 2.1).

Table 2.1: Correlation matrix for the variables of secondary schools (10 schools)

	<i>Total Attracted vehicles</i>	<i>Number of Students</i>	<i>Number of Class- rooms</i>	<i>Number of Employees</i>	<i>Lot area (per 1000 sqm)</i>	<i>Floor area (per 1000 sqm)</i>	<i>Parking space</i>	<i>Dist to nearest school m</i>	<i>Dist to highway m</i>	<i>Dist to collector m</i>	<i>Dist to CBD m</i>
Total Attracted	1.00										
Students	0.87	1.00									
Classes	0.76	0.94	1.00								
Employees	0.84	0.72	0.68	1.00							
Lot area	0.44	0.42	0.36	0.50	1.00						
Floor area	0.72	0.43	0.42	0.66	0.58	1.00					
Parking space area	0.50	0.43	0.47	0.33	0.36	0.35	1.00				
Dist to nearest school	0.01	-0.26	-0.34	-0.40	-0.23	0.23	0.12	1.00			
Dist to highway	0.14	0.02	-0.08	0.02	0.24	0.11	0.21	0.34	1.00		
Dist to collector	-0.29	-0.57	-0.66	-0.59	-0.27	0.00	0.01	0.91	0.35	1.00	
Dist to CBD	-0.12	-0.30	-0.34	-0.57	-0.26	0.04	0.25	0.92	0.20	0.89	1.00

From the correlation matrix, it is clear that the most significant variables which have a strong linear relationship with the dependent variable are the number of students, classrooms, employees and floor area.

The relationships between the dependent variable y (total attracted vehicles) and independent variables for secondary schools are summarized in Tables 2.2 to 2.5 as shown below.

Table 2.2: Summary of the developed models with intercept for 7 secondary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.2716*Students+67.479	0.84	780
Attracted vehicles = 4.8691*Employees-5.1964	0.62	1894
Attracted vehicles = 9.8685*Classes+54.793	0.75	1252
Attracted vehicles = 27.971*Floor area+44.14	0.39	3081
Attracted vehicles = 8.05*Lot area+101.96	0.05	4752
Attracted vehicles = -0.0045*Distance to nearest school+206.86	-0.19	5972
Attracted vehicles = -0.0121*Distance to nearest arterial+201.91	-0.20	6009
Attracted vehicles = -0.2385*Distance to nearest collector+227.4	0.098	4534
Attracted vehicles = -0.0026*Distance to CBD+206.29	-0.19	5977
Attracted vehicles = 7.48+0.226*Students+14.77*Floor area	0.97	132

Table 2.3: Summary of the developed models without intercept for 7 secondary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.3855*Students	0.80	1585
Attracted vehicles = 4.7539*Employees	0.80	1580
Attracted vehicles = 13.073*Classes	0.80	1532
Attracted vehicles = 35.26*Floor area	0.78	2749
Attracted vehicles = 15.703*Lot area	0.74	5014
Attracted vehicles = 0.0665*Distance to nearest school	0.27	28805
Attracted vehicles = 0.2862* Distance to nearest arterial	0.34	25135
Attracted vehicles = 0.5778*Distance to nearest collector	0.075	39215
Attracted vehicles = 0.0398*Distance to CBD	0.256	29829
Attracted vehicles = 0.228*Students+15.82*Floor area	0.80	112

Table 2.4: Summary of the developed models with intercept for 10 secondary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.2558*Students+80.745	0.72	1103
Attracted vehicles = 4.7015*Employees+6.19	0.67	1309
Attracted vehicles = 8.3744*Classes+83.546	0.52	1903
Attracted vehicles = 27.67*Floor area+50.472	0.46	2122
Attracted vehicles = 6.71*Lot area+124.56	0.10	3558
Attracted vehicles = 0.0005*Distance to nearest school+195.70	-0.12	4431
Attracted vehicles = 0.0236*Distance to nearest arterial+186.35	-0.10	4346
Attracted vehicles = -0.1097*Distance to nearest collector+213.61	-0.03	4069
Attracted vehicles = -0.0027*Distance to CBD+204.91	-0.11	4370
Attracted vehicles = 18.76+0.201*Students+16.41*Floor area	0.87	503

Table 2.5: Summary of the developed models without intercept for 10 secondary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.4046*Students	0.84	2180
Attracted vehicles = 4.8445*Employees	0.86	1166
Attracted vehicles = 13.717*Classes	0.83	2752
Attracted vehicles = 36.47*Floor area	0.84	2112
Attracted vehicles = 16.94*Lot area	0.77	5210
Attracted vehicles = 0.061*Distance to nearest school	0.41	22639
Attracted vehicles = 0.284*Distance to nearest arterial	0.48	19069
Attracted vehicles = 0.5776*Distance to nearest collector	0.27	29016
Attracted vehicles Y = 0.0362*Distance to CBD	0.39	23391
Attracted vehicles = 0.208*Students+19.12*Floor area	0.86	473

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easier to compare with the international models which use ratios in most situations.

2.1.2.2 Models Validation

The remaining three schools which were not included in developing the models of seven schools without intercept were used to validate these models. Table 2.6 includes the observed values of dependent variable in the first column and the predicted values by the developed models of seven schools without intercept in the other columns.

Table 2.6: Predicted values by different variables using the developed models (Secondary, Trip Attraction)

Number of attracted vehicles to school	Students (difference)	Employees (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)	Students + Floor Area (difference)
Coefficient	0.39	4.75	13.10	35.26	15.7	0.228, 15.82
245	173(-29%)	228(-7%)	131(-47%)	204(-17%)	145(-41%)	194(-21%)
172	97(-44%)	142(-17%)	118(-31%)	169(-2%)	83(+52)	133(-22%)
149	159(+7%)	152(+2%)	170(+14%)	114(-23%)	119(-20%)	146(-2%)
Average difference %	27	8.5	26	14	38	15
Range	7-44	1-18	16-38	2-23	20-52	0-22

CV: coefficient of variation = Standard deviation / Mean

$$\text{Difference} = \frac{\text{Predicted value} - \text{Observed value}}{\text{Observed value}} * 100\%$$

2.1.2.3 Discussion

By comparing the models which resulted from the seven schools regression with those that resulted from ten schools analysis, it can be concluded that the developed models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of ten schools analysis. In addition to that, there is no significant difference in the coefficients of independent variables between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of attracted vehicular trips to school when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 2.5 which summarized the results of analysis of ten schools without intercept and Table 2.6 which contains the predicted numbers of attracted vehicles using the developed models, the best multiple and simple regression models with the lowest mean square residual that explains the attraction behavior to secondary schools are as follows:

Model A: number of attracted vehicles = $0.208 * \text{Students} + 19.12 * \text{Floor Area}$

Model B: number of attracted vehicles = $4.84 * \text{Number of Employees}$

The models have the smallest mean square residual. The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the number of students, the plot of residuals against the total floor area, and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of attracted trips to secondary schools were arranged in order (from the smallest mean square error to the largest one) as follows:

Model C: number of attracted vehicles = 36.47 * Total Floor area

Model D: number of attracted vehicles = 0.40 * Number of Students

Model E: number of attracted vehicles = 13.72 * Number of Classes

Model F: number of attracted vehicles = 16.94 * Total Lot Area

The characteristics of these models are summarized in Table 2.7.

Table 2.7: Characteristics of the developed models (Secondary, Trip Attraction)

Model number	Variable(s)	Coefficients	P-Value	Mean square residual
A	Students	0.208	4.7E-04	473
	Floor Area	19.12	4.1E-04	
B	Employees	4.84	1.58E-08	1166
C	Floor Area	36.47	2.30E-07	2112
D	Students	0.40	2.66E-07	2188
E	Classes	13.72	7.62E-07	5210
F	Lot Area	16.94	3.0E-04	2752

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the independent variables and the normality plot of residuals. All assumptions seemed to be reasonable.

It is obvious that the worst models are those models that depend on the accessibility factors, such as distance to nearest school, nearest arterial, nearest collector, and distance to CBD since they have the highest mean square residual and lowest F-value, which means that the accessibility factors are not significant in determining the number of attracted vehicles to schools.

2.1.3 Data Analysis for Intermediary Schools

2.1.3.1 Summary of Results

The correlation between the dependent variable (total attracted vehicles) and independent variables for intermediary schools was introduced by the following matrix (Table 2.8).

Table 2.8: Correlation matrix for the variables of intermediary schools (10 schools)

	<i>Total Attracted vehicles</i>	<i>Number of Students</i>	<i>Number of Class- rooms</i>	<i>Number of Employees</i>	<i>Lot area (per 1000 sqm)</i>	<i>Floor area (per 1000 sqm)</i>	<i>Parking space</i>	<i>Dist to nearest school m</i>	<i>Dist to highway m</i>	<i>Dist to collector m</i>	<i>Dist to CBD m</i>
Total Attracted	1.00										
Students	0.94	1.00									
Classrooms	0.64	0.73	1.00								
Employees	0.88	0.91	0.65	1.00							
Lot area	0.83	0.76	0.25	0.85	1.00						
Floor area	0.91	0.89	0.64	0.97	0.83	1.00					
Parking space area	0.49	0.54	0.09	0.74	0.80	0.66	1.00				
Dist to nearest school	0.17	0.30	0.17	0.08	0.18	0.04	0.02	1.00			
Dist to highway	0.16	0.17	-0.16	0.05	0.17	0.02	0.32	-0.21	1.00		
Dist to collector	0.24	0.19	-0.27	-0.06	0.18	-0.01	-0.05	0.27	0.44	1.00	
Dist to CBD	0.09	0.11	-0.19	-0.18	0.01	-0.21	-0.10	0.26	0.65	0.86	1.00

From the correlation matrix, it is clear that the most significant variables which have a strong linear relationship with the dependent variable are the number of students, classrooms, employees, lot area and floor area.

The relationships between the dependent variable y (total attracted vehicles) and independent variables for intermediary schools are summarized in Tables 2.9 to 2.12 as shown below.

Table 2.9: Summary of the developed models with intercept for 7 intermediary schools (Trip Attraction)

Model	Adjusted R^2	Mean square residual
Attracted vehicles = $0.761 \times \text{Students} - 91.495$	0.88	1137
Attracted vehicles = $8.2 \times \text{Employees} - 101.71$	0.81	1798
Attracted vehicles = $15.95 \times \text{Classes} + 54.793$	0.30	6705
Attracted vehicles = $56.076 \times \text{Floor area} + 1.70$	0.86	1395
Attracted vehicles = $13.50 \times \text{Lot area} + 85.98$	0.62	3657
Attracted vehicles = $0.0235 \times \text{Distance to nearest school} + 204.31$	-0.18	11354
Attracted vehicles = $0.21 \times \text{Distance to nearest arterial} + 155.75$	0.04	9243
Attracted vehicles = $0.51 \times \text{Distance to nearest collector} + 177.31$	0.25	7212
Attracted vehicles = $0.011 \times \text{Distance to CBD} + 199.87$	-0.09	10504
Attracted vehicles = $-118.74 + 13.77 \times \text{Classes} + 12.44 \times \text{Lot area}$	0.988	111

Table 2.10: Summary of the developed models without intercept for 7 intermediary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.561*Students	0.81	1592
Attracted vehicles = 5.85*Employees	0.80	2205
Attracted vehicles = 14.895*Classes	0.76	5607
Attracted vehicles = 56.439*Floor area	0.82	1163
Attracted vehicles = 19.757*Lot area	0.77	4831
Attracted vehicles = 0.164*Distance to nearest school	0.60	16830
Attracted vehicles = 0.54*Distance to nearest arterial	0.64	13833
Attracted vehicles = 1.31*Distance to nearest collector	0.50	24320
Attracted vehicles = 0.053*Distance to CBD	0.45	27658
Attracted vehicles = 13.049*Classes+14.84*Lot area-6.98* Employees+36.94*Floor area	0.66	296

Table 2.11: Summary of the developed models with intercept for 10 intermediary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.7114*Students-65.292	0.87	923
Attracted vehicles = 6.73*Employees-34.34	0.75	1778
Attracted vehicles = 11.787*Classes+57.562	0.33	4670
Attracted vehicles = 49.339*Floor area+33.462	0.80	1413
Attracted vehicles = 13.226*Lot area+83.289	0.65	2474
Attracted vehicles = 0.028*Distance to nearest school+189.85	-0.09	7650
Attracted vehicles = 0.044*Distance to nearest arterial+199.43	-0.10	7677
Attracted vehicles = -0.1462*Distance to nearest collector+200.71	-0.06	7438
Attracted vehicles = -0.0028*Distance to CBD+211.69	-0.11	7812

Table 2.12: Summary of the developed models without intercept for 10 intermediary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.5597*Students	0.87	1120
Attracted vehicles = 5.8891*Employees	0.86	1671
Attracted vehicles = 15.57*Classes	0.82	4473
Attracted vehicles = 57*Floor area	0.87	1415
Attracted vehicles = 19.72*Lot area	0.83	3624
Attracted vehicles = 0.17*Distance to nearest school	0.68	13066
Attracted vehicles = 0.34*Distance to nearest arterial	0.60	17985
Attracted vehicles = 0.920*Distance to nearest collector	0.45	26867
Attracted vehicles = 0.041*Distance to CBD	0.48	25343

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

2.1.3.2 Models Validation

The remaining three schools which were not included in developing the models of seven schools without intercept were used to validate these models. Table 2.13 includes the observed values of dependent variable in the first column and the predicted values by the developed models of seven schools without intercept in the other columns.

Table 2.13: Predicted values by different variables using the developed models (Intermediary, Trip Attraction)

Number of attracted vehicles to school	Students (difference)	Employees (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)	Classes + Lot Area + Employees + Floor Area (difference)
Coefficient	0.56	5.85	14.90	56.4	19.76	13,14.84,6.98,36.94
160	157(-1%)	123(-23%)	104(-35%)	102(-36%)	104(-35%)	89(-44%)
199	218(10%)	217(+9%)	149(-25%)	231(+16%)	219(+10%)	188(-6%)
223	210(-6%)	211(-5%)	179(-20%)	186(-17%)	232(4%)	202(-9%)
Average difference %	6	12	27	23	16	19
Range %	1-10	5-23	20-35	16-36	4-35	6-44

CV: coefficient of variation = Standard deviation / Mean

$$\text{Difference} = \frac{\text{Predicted value} - \text{Observed value}}{\text{Observed value}} * 100\%$$

2.1.3.3 Discussion

By comparing the models which resulted from the seven schools analysis with those that resulted from the ten schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of the ten schools analysis. On the other

hand, there is no significant difference in the coefficients of independent variables between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of attracted vehicular trips to school when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

The result of multiple regression analysis is a simple regression model. Because of that, the best models are simple models.

Based on Table 2.12 which summarized the results of analysis of ten schools without intercept and Table 2.13 which contains the predicted numbers of attracted vehicles using the developed models, the best regression model with the lowest mean square residual that explains the attraction behavior to secondary schools is as follows:

Model A: number of attracted vehicles = 0.56 * Number of students

The model has the smallest mean square residual. The assumptions of normality, independency and equal variance of the model were checked graphically by considering the plot of residuals against the number of classes, the plot of residuals against the total lot area, the plot of residuals against the number of employees, the plot of residuals

against the total floor area, and the normality plot of residuals. All assumptions seemed to be reasonable as shown in appendix A.

The other models which are significant in predicting the number of attracted trips to intermediary schools were arranged in order (from the smallest mean square error to the largest one) as follows:

Model B: number of attracted vehicles = 57 * Total Floor Area

Model C: number of attracted vehicles = 5.89 * Number of Employees

Model D: number of attracted vehicles = 19.72 * Total Lot Area

Model E: number of attracted vehicles = 15.57 * Classes

The characteristics of these models are summarized in Table 2.14 as shown below.

Table 2.14: Characteristics of the developed models (Intermediary, Trip Attraction)

Model number	Variable	Coefficient	P-Value	Mean square residual
A	Students	0.56	3.9E-09	1120
B	Floor Area	57	1.11E-08	1415
C	Employees	5.89	2.36E-08	1671
D	Lot Area	19.72	7.78E-07	3624
E	Classes	15.57	2.02E-06	4473

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the number of students

and the normality plot of residuals. All assumptions seemed to be reasonable as shown in appendix A.

It is obvious that the worst models are those models that depend on the accessibility factors such as distance to nearest school, nearest arterial, nearest collector, and distance to CBD, which means that the accessibility factors are not significant in determining the number of attracted vehicles to schools.

2.1.4 Data Analysis for Primary Schools

2.1.4.1 Summary of Results

The correlation between the dependent variable (total attracted vehicles) and independent variables for primary schools was introduced by the following matrix (Table 2.15).

Table 2.15: Correlation matrix for the variables of primary schools (10 schools)

	<i>Total Attracted</i>	<i>Number of Students</i>	<i>Number of Class- rooms</i>	<i>Number of Employees</i>	<i>Lot area (per 1000 sqm)</i>	<i>Floor area (per 1000 sqm)</i>	<i>Parking space</i>	<i>Dist to nearest school m</i>	<i>Dist to arterial m</i>	<i>Dist to collector m</i>	<i>Dist to CBD m</i>
Total Attracted	1.00										
Students	0.85	1.00									
Classrooms	0.86	0.97	1.00								
Employees	0.88	0.92	0.97	1.00							
Plot area	0.78	0.42	0.52	0.63	1.00						
Floor area	0.87	0.78	0.83	0.86	0.80	1.00					
Parking space area	0.66	0.48	0.49	0.64	0.74	0.67	1.00				
Dist to nearest school	-0.46	-0.57	-0.56	-0.56	-0.26	-0.62	-0.45	1.00			
Dist to highway	0.36	0.04	0.08	0.17	0.47	0.38	0.41	-0.05	1.00		
Dist to collector	-0.02	-0.16	-0.17	-0.16	-0.02	-0.11	-0.06	0.58	0.65	1.00	
Dist to CBD	0.14	-0.11	-0.05	-0.01	0.33	0.01	0.03	0.70	0.50	0.74	1.00

From the correlation matrix, it is clear that the most significant variables which have a strong linear relationship with the dependent variable are the number of students, classrooms, employees, lot area and floor area.

The relationships between the dependent variable y (total attracted vehicles) and independent variables for primary schools are summarized in Tables 2.16 to 2.19 as shown below.

Table 2.16: Summary of the developed models with intercept for 7 primary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.3695*Students+41.654	0.68	1811
Attracted vehicles = 4.6515*Employees+24.866	0.70	1679
Attracted vehicles = 9.0665*Classes+31.241	0.70	1733
Attracted vehicles = 46.274*Floor area+57.136	0.70	1681
Attracted vehicles = 9.2706*Lot area+107.12	0.54	2621
Attracted vehicles = -0.038*Distance to nearest school+219.30	0.18	4678
Attracted vehicles = 0.16*Distance to nearest arterial+122.63	0.004	5656
Attracted vehicles = 0.51*Distance to nearest collector+177.31	0.25	7212
Attracted vehicles = -0.0001*Distance to CBD+174.11	-0.20	6820
Attracted vehicles = 33.7+0.272*Students+5.92*Lot area	0.90	550

Table 2.17: Summary of the developed models without intercept for 7 primary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.4662*Students	0.079	1854
Attracted vehicles = 5.3181*Employees	0.80	1502
Attracted vehicles = 10.753*Classes	0.79	1617
Attracted vehicles = 64.131*Floor area	0.78	2211
Attracted vehicles = 18.166*Lot area	0.65	7585
Attracted vehicles = 0.067*Distance to nearest school	0.16	27688
Attracted vehicles = 0.464*Distance to nearest arterial	0.61	8951
Attracted vehicles = 0.31*Distance to nearest collector	0.044	32244
Attracted vehicles = 0.028*Distance to CBD	0.38	18693
Attracted vehicles = 0.346*Students+6.20*Lot area	0.78	708

Table 2.18: Summary of the developed models with intercept for 10 primary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.3607*Students+43.64	0.69	1407
Attracted vehicles = 4.7976*Employees+21.534	0.74	1141
Attracted vehicles = 8.6752*Classes+38.969	0.70	1324
Attracted vehicles = 44.362*Floor area+54.562	0.72	1240
Attracted vehicles = 9.5437*Lot area+105.56	0.56	1976
Attracted vehicles = -0.032*Distance to nearest school+203.25	0.11	3966
Attracted vehicles = 0.066*Distance to nearest arterial+145.75	0.02	4364
Attracted vehicles = -0.005*Distance to nearest collector+172.84	-0.12	5028
Attracted vehicles = 0.003*Distance to CBD+ 160.22	-0.10	4927
Attracted vehicles = 32.59+0.269*Students+6.29*Lot area	0.92	359

Table 2.19: Summary of the developed models without intercept for 10 primary schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.4652*Students	0.85	1568
Attracted vehicles = 5.402*Employees	0.86	1077
Attracted vehicles = 10.855*Classes	0.85	1420
Attracted vehicles = 61.285*Floor area	0.84	1701
Attracted vehicles = 19.315*Lot area	0.72	6194
Attracted vehicles = 0.078*Distance to nearest school	0.23	24701
Attracted vehicles = 0.27*Distance to nearest arterial	0.50	14314
Attracted vehicles = 0.31*Distance to nearest collector	0.21	25347
Attracted vehicles = 0.031*Distance to CBD	0.48	15372
Attracted vehicles = 0.34*Students+6.67*lot Area	0.86	509

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

2.1.4.2 Models Validation

The remaining three schools which were not included in developing the models of seven schools without intercept were used to validate these models. Table 2.20 includes the observed values of dependent variable in the first column and the predicted values by the developed models of seven schools without intercept in the other columns.

Table 2.20: Predicted values by different variables using the developed models (Primary, Trip Attraction)

Number of attracted vehicles to school	Students (difference)	Employees (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)	Students + Lot Area (difference)
Coefficient	0.47	5.32	10.75	64.13	18.17	0.35,6.20
165	205(+24%)	170(+3%)	194(+18%)	201(+22%)	142(-14%)	176(+6.5%)
113	88(-22%)	106(+6%)	75(-34%)	102(-10%)	154(+36)	97(-14%)
223	196(-12%)	197(-12%)	194(-13%)	259(+16%)	201(-10%)	208(-7%)
Average difference %	19	7	22	16	20	9
Range%	12-24	3-12	13-34	10-22	10-36	6-14

CV: coefficient of variation = Standard deviation / Mean

$$\text{Difference} = \frac{\text{Predicted value} - \text{Observed value}}{\text{Observed value}} * 100\%$$

2.1.4.3 Discussion

By comparing the models which resulted from the seven schools analysis with those that resulted from the ten schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of the ten schools analysis. On the other

hand, there is no significant difference in the coefficients of independent variables between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of attracted vehicular trips to school when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 2.19 which summarized the results of analysis of ten schools without intercept and Table 2.20 which contains the predicted numbers of attracted vehicles using the developed models, the best multiple and simple regression models with the lowest mean square residual that explains the attraction behavior to secondary schools are as follows:

Model A: number of attracted vehicles = $0.34 * \text{Students} + 6.67 * \text{Lot Area}$

Model B: number of attracted vehicles = $5.40 * \text{Number of Employees}$

The models have the smallest mean square residual. The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the number of students, the plot of residuals against the total lot area, and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of attracted trips to primary schools were arranged in order (from the smallest mean square residual to the largest one) as follows:

Model C: number of attracted vehicles = 10.86 * Number of Classes

Model D: number of attracted vehicles = 0.465 * Number of Students

Model E: number of attracted vehicles = 61.29 * Total Floor Area

Model F: number of attracted vehicles =19.315 * Total Lot Area

The characteristics of these models are summarized in Table 2.21 as shown below.

Table 2.21: Characteristics of the developed models (Primary, Trip Attraction)

Model number	Variable	Coefficient	P-Value	Mean square residual
A	Students	0.34	8.06E-06	509
	Lot Area	6.67	2.0E-03	
B	Employees	5.40	3.12E-08	1077
C	Classes	10.86	1.09E-07	1420
D	Students	0.465	1.70E-07	1568
E	Floor Area	61.29	2.46E-07	1701
F	Lot Area	19.32	8.70E-05	6192

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the independent variables and the normality plot of residuals. All assumptions seemed to be reasonable.

It is obvious that the worst models are those models that depend on the accessibility factors such as distance to nearest school, nearest arterial, nearest collector, and distance to CBD, which means that the accessibility factors are not significant in determining the number of attracted vehicles to schools.

2.1.5 Data Analysis for All Levels Schools

2.1.5.1 Summary of Results

The simple regression analysis for all levels schools was done in the same procedure as each level of schools. The only difference is in multiple regression analysis which was done two times. The first time was done without distinguishing between the levels of schools. In the second multiple regression analysis, two dummy variables were used to distinguish between the levels of schools as follows:

Dummy 1 = 1 if secondary school, 0 otherwise

Dummy 2 = 1 if intermediary school, 0 otherwise

The correlation between the dependent variable (total attracted vehicles) and independent variables for all levels of schools was introduced by the following matrix (Table 2.22).

Table 2.22: Correlation matrix for the variables of all levels of schools (30 schools)

	<i>Total Attracted</i>	<i>Number of Students</i>	<i>Number of Class- rooms</i>	<i>Number of Employees</i>	<i>Lot area (per 1000 sqm)</i>	<i>Floor area (per 1000 sqm)</i>	<i>Parking space</i>	<i>Dist to nearest school ml</i>	<i>Dist to highway m</i>	<i>Dist to collector m</i>	<i>Dist to CBD m</i>
Total Attracted	1.00										
Students	0.80	1.00									
Classrooms	0.66	0.82	1.00								
Employees	0.84	0.81	0.69	1.00							
Lot area	0.72	0.51	0.31	0.70	1.00						
Floor area	0.71	0.63	0.40	0.80	0.72	1.00					
Parking space area	0.53	0.47	0.30	0.62	0.70	0.60	1.00				
Dist to nearest school	-0.06	-0.19	-0.34	-0.22	-0.06	0.16	0.00	1.00			
Dist to highway	0.24	0.07	-0.04	0.10	0.32	0.16	0.31	0.13	1.00		
Dist to collector	-0.05	-0.26	-0.30	-0.27	-0.07	-0.09	-0.06	0.55	0.48	1.00	
Dist to CBD	0.04	-0.15	-0.17	-0.24	0.04	-0.09	0.01	0.61	0.43	0.77	1.00

From the correlation matrix, it is clear that the most significant variables which have a strong linear relationship with the dependent variable are the number of students, employees, lot area and floor area.

The relationships between the dependent variable y (total attracted vehicles) and independent variables for intermediary schools are summarized in Tables 2.23 to 2.26 as shown below.

Table 2.23: Summary of the developed models with intercept for 21 pooled schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.3566*Students+51.076	0.63	2464
Attracted vehicles = 5.4388*Employees-6.2025	0.69	2073
Attracted vehicles = 10.33*Classes+43.399	0.49	3400
Attracted vehicles = 28.714*Floor area+85.338	0.46	3620
Attracted vehicles = 10.262*Lot area+98.791	0.49	3454
Attracted vehicles = -0.012*Distance to nearest school+218.57	-0.02	6880
Attracted vehicles = 0.075*Distance to nearest arterial+176.8	-0.0017	6735
Attracted vehicles = -0.077*Distance to nearest collector+212.5	-0.026	6900
Attracted vehicles = -0.0009*Distance to CBD+199.47	-0.05	7071
Attracted vehicles = -2.05+8.06*Classes+7.96*Lot area	0.78	1472
Attracted vehicles = 21.87+0.237*Students +16.08*Floor area -74.53*Dummy1+3.9*Lot area	0.88	757

Table 2.24: Summary of the developed models without intercept for 21 pooled schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.458*Students	0.89	2763
Attracted vehicles = 5.292*Employees	0.91	1973
Attracted vehicles = 12.823*Classes	0.88	3460
Attracted vehicles = 45.766*Floor area	0.85	4863
Attracted vehicles = 17.761*Lot area	0.84	5683
Attracted vehicles = 0.0867*Distance to nearest school	0.41	26735
Attracted vehicles = 0.41*Distance to nearest arterial	0.61	16997
Attracted vehicles = 0.552*Distance to nearest collector	0.24	35333
Attracted vehicles = 0.037*Distance to CBD	0.45	24830
Attracted vehicles = 7.97*Classes+7.92*Lot area	0.92	1395
Attracted vehicles = 0.264*Students +17.52*Floor area-76.52*Dummy1+4.10*Lot area	0.93	794

Table 2.25: Summary of the developed models with intercept for 30 pooled schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.3487*Students+55.817	0.62	1976
Attracted vehicles = 5.4388*Employees-6.2025	0.70	1575
Attracted vehicles = 8.6586*Classes+73.326	0.42	3026
Attracted vehicles = 28.088*Floor area+86.737	0.48	2697
Attracted vehicles = 10.188*Lot area+101.12	0.51	2578
Attracted vehicles = -0.0039*Distance to nearest school+201.69	-0.03	5374
Attracted vehicles = 0.051*Distance to nearest arterial+173.87	0.02	5081
Attracted vehicles = -0.019*Distance to nearest collector+199.74	-0.03	5382
Attracted vehicles = 0.0012*Distance to CBD+ 192.67	-0.03	5386
Attracted vehicles = 37.65+0.253*Students+6.05*Lot area	0.75	1289
Attracted vehicles = 30.85+0.21*Students +18.23*Floor area -59.39*Dummy1+3.13*Lot area	0.85	801

Table 2.26: Summary of the developed models without intercept for 30 pooled schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.4678*Students	0.91	2355
Attracted vehicles = 5.2923*Employees	0.93	1526
Attracted vehicles = 13.155*Classes	0.89	3629
Attracted vehicles = 46.437*Floor area	0.88	3949
Attracted vehicles = 18.564*Lot area	0.86	4860
Attracted vehicles = 0.082*Distance to nearest school	0.44	23724
Attracted vehicles = 0.30*Distance to nearest arterial	0.61	16281
Attracted vehicles = 0.49*Distance to nearest collector	0.32	28936
Attracted vehicles = 0.035*Distance to CBD	0.52	20255
Attracted vehicles = 0.316*Students+6.83*Lot area	0.93	1435
Attracted vehicles = 0.252*Students +20.51*Floor Area- 61.79*Dummy1+3.34*Lot area	0.94	908

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of 30 schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

2.1.5.2 Models Validation

The remaining nine schools which were not included in developing the models were used to validate these models. Table 2.27 includes the observed values of dependent variable in the first column and the predicted values by all the developed models in the other columns.

Table 2.27: Predicted values by different variables using the developed models (All Levels, Trip Attraction)

Number of attracted vehicles to school	Employees (difference)	Students (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)	Classes + Lot Area (difference)	Students + Dummy1 + Lot Area + Floor Area (difference)
Coefficients	5.292	0.49	12.82	45.77	17.76	7.97,7.92	0.264,17.52,-76.52,4.1
245	254(+3.5%)	205(-16%)	128(-47%)	264(+7%)	164(-33%)	153(-38%)	181(-26%)
172	159(-7.5%)	115(-33%)	115(-33%)	220(+28%)	94(-46%)	113(-34%)	95(-44%)
149	169(+13.5)	199(+33%)	167(+12%)	148(0%)	135(-9%)	164(+10%)	121(-19%)
160	111(-30.5%)	128(-20%)	90(-44%)	82(-49%)	94(-42%)	97(-39%)	127(-20.5%)
199	196(-1.5%)	178(-11%)	128(-36%)	188(-6%)	197(-1%)	168(-16%)	220(+10%)
223	191(-14.5%)	172(-23%)	154(-31%)	151(-32%)	209(-6%)	189(-15%)	205(-8%)
165	169(+2.5%)	202(+22%)	231(+40%)	144(-13%)	67(-59%)	173(+5%)	186(+13%)
113	106(-6%)	86(-24%)	90(-21%)	73(+36%)	91(-20%)	96(-15%)	98(-13%)
223	196(-12%)	192(-14%)	231(+3.5%)	185(-17%)	180(-19%)	224(0%)	223(0%)
Average difference %	10	20	29	21	26	19	17
Range	1.5-30.5	11-33	3.5-47	0-49	1-59	0-39	0-44

CV: coefficient of variation = Standard deviation / Mean

$$\text{Difference} = \frac{\text{Predicted value} - \text{Observed value}}{\text{Observed value}} * 100\%$$

2.1.5.3 Discussion

By comparing the models which resulted from the 21 schools analysis with those that resulted from the 30 schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of 30 schools analysis. On the other hand, there is no significant difference in the coefficients of independent variables between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of attracted vehicular trips to school when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors”. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 2.26 which summarized the results of analysis of 30 schools without intercept and Table 2.27 which contains the predicted numbers of attracted vehicles using the developed models, the best multiple and simple regression models with the lowest mean square residual that explains the attraction behavior to secondary schools are as follows:

$$\begin{aligned} \text{Model A: number of attracted vehicles} &= 0.252 * \text{Number of Students} \\ &+ 20.51 * \text{Floor Area} - 61.79 * \text{Dummy} + 3.34 * \text{Lot Area} \end{aligned}$$

Where, Dummy1 = 1 if secondary, 0 otherwise

Model B: number of attracted vehicles = 5.29 * Number of Employees

The models have the smallest mean square residual. The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the number of employees and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of attracted trips to schools were arranged in order (from the smallest mean square error to the largest one) as follows:

Model C: number of attracted vehicles = 0.316*Students+6.83*Lot Area

Model D: number of attracted vehicles = 0.468 * Number of Students

Model E: number of attracted vehicles = 13.16 * Number of Classes

Model F: number of attracted vehicles = 46.44 * Total Floor Area

Model G: number of attracted vehicles = 18.56 * Total Lot Area

The characteristics of these models are summarized in Table 2.28 as shown below.

Table 2.28: Characteristics of the developed models (All Levels, Trip Attraction)

Model number	Variable	Coefficient	P-Value	Mean square residual
A	Students	0.252	7.7E-07	908
	Floor Area	20.51	2.0E-03	
	Lot Area	3.34	5.0E-02	
	Dummy1	61.79	3.0E-04	
B	Employees	5.29	6.89E-23	1526
C	Students	0.316	4.12E-09	1435
	Lot Area	6.83	1.0E-04	
D	Students	0.468	3.75E-20	2355
E	Classes	13.16	2.0E-17	3629
F	Floor Area	46.44	6.88E-17	3949
G	Lot Area	18.56	1.41E-15	4860

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the independent variables and the normality plot of residuals. All assumptions seemed to be reasonable.

It is obvious that the worst models are those models that depend on the accessibility factors, such as distance to nearest school, nearest arterial, nearest collector, and distance to CBD, which means that the accessibility factors are not significant in determining the number of attracted vehicles to schools.

From “**model A**”, it is clear that the secondary schools have the significant effect because of the appearance of the dummy variable which is related to secondary schools. Because of that, another trial of analysis in which the primary and intermediary schools were pooled together, was performed as shown in the next section.

2.1.6 Data Analysis for Primary & Intermediary Schools

2.1.6.1 Summary of Results

The correlation between the dependent variable (total attracted vehicles) and independent variables for primary and intermediary schools was introduced by the following matrix (Table 2.29).

Table 2.29: Correlation matrix for the variables of primary & intermediary schools (20 schools)

	<i>Total Attracted vehicles</i>	<i>Number of Students</i>	<i>Number of Classes</i>	<i>Number of Employees</i>	<i>Lot area (per 1000 sqm)</i>	<i>Floor area (per 1000 sqm)</i>	<i>Parking space</i>	<i>Dist to nearest school m</i>	<i>Dist to highway m</i>	<i>Dist to collector m</i>	<i>Dist to CBD m</i>
Total Attracted	1.00										
Students	0.85	1.00									
Classes	0.64	0.85	1.00								
Employees	0.88	0.91	0.76	1.00							
Lot area	0.82	0.57	0.34	0.76	1.00						
Floor area	0.90	0.80	0.60	0.91	0.83	1.00					
Parking space area	0.56	0.51	0.31	0.68	0.76	0.65	1.00				
Dist to nearest school	-0.15	-0.31	-0.37	-0.30	-0.07	-0.28	-0.27	1.00			
Dist to highway	0.29	0.11	-0.03	0.16	0.36	0.24	0.38	-0.08	1.00		
Dist to collector	0.02	-0.10	-0.17	-0.16	-0.01	-0.12	-0.07	0.50	0.54	1.00	
Dist to CBD	0.10	-0.03	-0.10	-0.09	0.16	-0.10	-0.04	0.53	0.55	0.74	1.00

From the correlation matrix, it is clear that the most significant variables which have a strong linear relationship with the dependent variable are the number of students, employees, lot area and floor area.

The relationships between the dependent variable y (total attracted vehicles) and independent variables for primary and intermediary schools pooled together are summarized in Tables 2.30 to 2.33 as shown below.

Table 2.30: Summary of the developed models with intercept for 14 schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.52*Students-0.086	0.72	2171
Attracted vehicles = 6.09*Employees-18.28	0.70	1930
Attracted vehicles = 10.69*Classes+57.40	0.40	4834
Attracted vehicles = 48.27*Floor area+43.02	0.82	1464
Attracted vehicles = 11.71*Lot area+97.56	0.64	2868
Attracted vehicles = -0.02*Distance to nearest school+233.67	0.02	8208
Attracted vehicles = 0.21*Distance to nearest arterial+132.17	0.14	6891
Attracted vehicles = 0.03*Distance to nearest collector+207.65	0.08	8656
Attracted vehicles = 0.002*Distance to CBD+196	0.08	8647
Attracted vehicles = 4.40+0.35*Students+6.73*Lot area	0.88	999

Table 2.31: Summary of the developed models without intercept for 14 schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.519*Students	0.88	2004
Attracted vehicles = 5.63*Employees	0.89	1820
Attracted vehicles = 12.71*Classes	0.836	4604
Attracted vehicles = 58.70*Floor area	0.89	1738
Attracted vehicles = 19.17*Lot area	0.81	5805
Attracted vehicles = 0.11*Distance to nearest school	0.43	25852
Attracted vehicles = 0.51*Distance to nearest arterial	0.72	10742
Attracted vehicles = 0.54*Distance to nearest collector	0.23	36244
Attracted vehicles = 0.037*Distance to CBD	0.45	24384
Attracted vehicles = 0.36*Students+6.73*Lot area	0.90	918

Table 2.32: Summary of the developed models with intercept for 20 schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.50*Students+8.83	0.72	1731
Attracted vehicles = 5.80*Employees-4.56	0.76	1457
Attracted vehicles = 8.92*Classes+66.36	0.37	3824
Attracted vehicles = 46.51*Floor area+46.57	0.80	1199
Attracted vehicles = 11.61*Lot area+95.67	0.66	2099
Attracted vehicles = -0.016*Distance to nearest school+213.24	-0.03	6274
Attracted vehicles = 0.067*Distance to nearest arterial+166.50	0.03	5889
Attracted vehicles = 0.005*Distance to nearest collector+195.57	-0.05	6423
Attracted vehicles = 0.003*Distance to CBD+ 186.36	-0.04	6355
Attracted vehicles = 10+0.33*Students+6.99*Lot area	0.88	715

Table 2.33: Summary of the developed models without intercept for 20 schools (Trip Attraction)

Model	Adjusted R²	Mean square residual
Attracted vehicles = 0.52*Students	0.91	1649
Attracted vehicles = 5.69*Employees	0.92	1382
Attracted vehicles = 12.91*Classes	0.86	4185
Attracted vehicles = 58.45*Floor area	0.91	1530
Attracted vehicles = 19.58*Lot area	0.85	4655
Attracted vehicles = 0.12*Distance to nearest school	0.48	21607
Attracted vehicles = 0.31*Distance to nearest arterial	0.61	15736
Attracted vehicles = 0.46*Distance to nearest collector	0.30	30165
Attracted vehicles = 0.036*Distance to CBD	0.52	19833
Attracted vehicles = 0.36*Students+6.98*Lot area	0.93	688

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

2.1.6.2 Models Validation

The remaining six schools which were not included in developing the models of 14 schools without intercept were used to validate these models. Table 2.34 includes the observed values of dependent variable in the first column and the predicted values by the developed models of seven schools without intercept in the other columns.

Table 2.34: Predicted values by the developed models (Primary & Intermediary, Trip Attraction)

Observed values	Students (difference)	Employees (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)	Students + Lot Area (difference)
Coefficient	0.52	5.63	12.71	58.70	19.17	0.36,6.73
165	228(+38%)	182(+10%)	232(+41%)	183(+11%)	74(-55%)	184(+12%)
113	98(-12%)	114(0%)	90(-20%)	93(-18%)	100(-11%)	102(-10%)
223	218(-2%)	211(-5%)	232(+4%)	236(+6%)	199(-11%)	219(-2%)
160	146(-9%)	120(-25%)	90(-44%)	105(-34%)	103(-36%)	136(-15%)
199	202(+1%)	211(+6%)	129(-35%)	239(+20%)	217(+10%)	215(+8%)
223	195(-12%)	205(-8%)	155(-30%)	193(-13%)	230(+3%)	214(-4%)
Average difference %	12	9	29	17	21	8.5
Range%	1-38	0-10	4-44	6-34	3-55	2-15

CV: coefficient of variation = Standard deviation / Mean

$$\text{Difference} = \frac{\text{Predicted value} - \text{Observed value}}{\text{Observed value}} * 100\%$$

2.1.6.3 Discussion

By comparing the models which resulted from the 14 schools analysis with those that resulted from the 20 schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of 20 schools analysis. On the other hand, there is no significant difference in the coefficients of independent variables between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of attracted vehicular trips to school when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 2.33 which summarized the results of analysis of ten schools without intercept and Table 2.34 which contains the predicted numbers of attracted vehicles using the developed models, the best multiple and simple regression models with the lowest mean square residual that explains the attraction behavior to secondary schools are as follows:

Model A: number of attracted vehicles = 0.36 * Students + 6.98 * Lot Area

Model B: number of attracted vehicles = 5.69 * Number of Employees

The models have the smallest mean square residual. The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the number of students, the plot of residuals against the total lot area, and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of attracted trips to primary and intermediary schools were arranged in order (from the smallest mean square error to the largest one) as follows:

Model C: number of attracted vehicles = 58.48 * Total Floor Area

Model D: number of attracted vehicles = 0.52 * Number of Students

Model E: number of attracted vehicles = 12.91 * Number of Classes

Model F: number of attracted vehicles = 19.58 * Total Lot Area

The characteristics of these models are summarized in Table 2.35 as shown below.

Table 2.35: Characteristics of the developed models (Primary & Intermediary, Trip Attraction)

Model number	Variable	Coefficient	P-Value	Mean square residual
A	Students	0.36	4.11E-09	688
	Lot area	6.98	5.45E-05	
B	Employees	5.69	5.45E-16	1382
C	Floor area	58.48	1.43E-15	1530
D	Students	0.52	2.920E-15	1649
E	Classes	12.91	2.08E-11	4185
F	Lot area	19.58	5.750E-11	4655

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the independent variables and the normality plot of residuals. All assumptions seemed to be reasonable.

It is obvious that the worst models are those models that depend on the accessibility factors, such as distance to nearest school, nearest arterial, nearest collector, and distance to CBD, which means that the accessibility factors are not significant in determining the number of attracted vehicles to schools.

2.2 Comparison between Local & International Models

The comparison was done between the trip generation manual of Riyadh city which presents the only option for local models and the international models such as the Institute of Transportation Engineers (ITE) trip generation manual and Dubai trip generation and parking rates manual. Tables 2.36 to 2.44 present the predicted values by all of these models for different variables for each level of schools.

2.2.1 Secondary Schools

Table 2.36: Predicted values by ITE rates for secondary schools (Trip Attraction)

Number of attracted vehicles	Students (difference)	Employees (difference)	1000 sq m Gross Floor Area (difference)
290	363(+25%)	239(-18%)	178(-39%)
240	303(+26%)	211(-12%)	188(-22%)
250	226(-10%)	257(+3%)	277(+11%)
157	130(-17%)	112(-28%)	192(+22.5)
235	210(-10%)	248(+6%)	224(-5%)
121	123(+2%)	173(+43%)	106(-12%)
107	80(-25%)	140(+31%)	120(+12)
245	188(-23%)	224(-8%)	190(-22%)
172	105(-39%)	140(-18%)	158(-8%)
149	174(+17%)	149(0%)	106(-28.5%)
Average difference %	19.4	16.7	18.2
Range of difference	2-39	0-43	5-39

Table 2.37: Predicted values by Riyadh city manual for secondary schools (Trip Attraction)

Number of attracted vehicles	Students (difference)	Lot Area (difference)
290	565(+95%)	1202(+314%)
240	471(+96%)	1653(+589%)
250	351(+40%)	1334(+434%)
157	203(+29%)	1543(+883%)
235	327(+39%)	2246(+855%)
121	191(+58%)	1160(+859%)
107	125(+17%)	746(+597%)
245	293(+20%)	1073(+380%)
172	172(-4.5%)	611(+255%)
149	271(+82%)	882(+492%)
Average difference %	48	566
Range of difference	4.5-96	255-883

Table 2.38: Predicted values by Dubai trip generation manual for secondary schools (Trip Attraction)

Number of attracted vehicles	Students (difference)
290	449(+55%)
240	374(+56%)
250	279(+11%)
157	161(+2.5%)
235	260(+10.5%)
121	152(+26%)
107	99(-7%)
245	233(-5%)
172	131(-24%)
149	215(+44%)
Average difference %	24
Range of difference	5-56

2.2.2 Intermediary Schools

Table 2.39: Predicted values by ITE rates for intermediary schools (Trip Attraction)

Number of Attracted vehicles	Students (difference)	Employees (difference)	1000 sq m Gross Floor Area (difference)
333	276(-17%)	276(-17%)	266(-20%)
321	319(-1%)	254(-21%)	229(-28.5%)
261	228(-13%)	244(-7%)	250(-4%)
270	254(-6%)	212(-21%)	208(-23%)
240	225(-6%)	223(-7%)	197(-18%)
150	202(+35%)	212(+41%)	167(+12%)
57	108(+89)	95(+67%)	36(-37%)
160	151(-5.5%)	111(-30%)	84(-47%)
199	210(+5.5%)	196(-1.5%)	192(-3.5%)
223	202(-9%)	191(-14%)	155(-31%)
Average difference %	19	23	22
Range of difference	1-89	7-67	3.5-47
CV	0.26	0.27	0.38

Table 2.40: Predicted values by Riyadh city manual for intermediary schools (Trip Attraction)

Number of attracted vehicles	Students (difference)	Gross Floor Area (difference)	Classes (difference)
333	384(+15%)	1306(+293%)	345(+4%)
321	443(+39%)	1127(+251%)	415(+29%)
261	317(+21%)	1226(+370%)	345(+32%)
270	353(+31%)	1021(+278%)	530(+96%)
240	313(+30%)	966(+303%)	345(+44%)
150	281(+88%)	822(+448%)	322(+114%)
57	150(+160%)	176(+209%)	230(+304%)
160	210(+31%)	414(+159%)	161(0%)
199	292(+47%)	943(+374%)	230(+16%)
223	281(+26%)	759(+240%)	276(+24%)
Average difference %	49	293	66
Range of difference	15-160	159-448	0-114

Table 2.41: Predicted values by Dubai trip generation manual for intermediary schools (Trip Attraction)

Number of attracted vehicles	Students (difference)
333	220(-34%)
321	254(-21%)
261	181(-30%)
270	202(-25%)
240	179(-25%)
150	161(+7.5%)
57	86(+51%)
160	120(-25%)
199	167(-16%)
223	161(-28%)
Average difference %	26
Range of difference	7.5-51

2.2.3 Primary Schools

Table 2.42: Predicted values by ITE rates for primary schools (Trip Attraction)

Number of attracted vehicles	Students (difference)	Employees (difference)	1000 sq m Gross Floor Area (difference)
260	185(-29%)	231(-11%)	234(-10%)
170	194(+14%)	242(+42%)	185(+8.5%)
220	169(-23%)	166(-24%)	176(-21%)
156	140(-10%)	150(-3.5%)	89(-43%)
250	296(+18%)	258(+3%)	205(-18%)
100	91(-9%)	91(-9%)	16(-84%)
60	51(-14%)	64(+7%)	83(+39%)
165	198(+20%)	171(+4%)	176(+6%)
113	85(-25%)	107(-5%)	89(-21%)
223	189(15%)	199(-11%)	226(+1.5%)
Average difference %	18	12	25
Range of difference	9-29	3-42	1.5-84

Table 2.43: Predicted values by Riyadh city manual for primary schools (Trip Attraction)

Number of attracted vehicles	Students (difference)	Lot Area (difference)	Employees (difference)
260	271(+4%)	1859(+615%)	428(+65%)
170	284(+67%)	823(+384%)	448(+164%)
220	248(+12.5%)	1166(+430%)	309(+40%)
156	205(+32%)	230(+47%)	279(+79%)
250	434(+74%)	664(+165%)	478(+91%)
100	134(+34%)	28(-72%)	169(+69%)
60	76(+26.5%)	258(+330%)	120(+99%)
165	290(+76%)	377(+128%)	319(+93%)
113	124(+10%)	511(+352%)	199(+76%)
223	277(+24%)	1014(+355%)	369(+65%)
Average difference %	36	288	84
Range of difference	4-74	47-615	40-164

Table 2.44: Predicted values by Dubai trip generation manual for primary schools (Trip Attraction)

Number of attracted vehicles	Students (difference)
260	176(-32%)
170	185(+9%)
220	161(-27%)
156	134(-14%)
250	283(+13%)
100	87(-13%)
60	49(-18%)
165	189(+15%)
113	81(-28%)
223	181(-19%)
Average difference %	19
Range of difference	9-32

2.3 Summary of Results

Tables 2.45 to 2.49 present the best models developed by the study for each type of analysis.

Table 2.45: Summary of the developed models for secondary schools (Trip Attraction)

Model number	Developed models	Adjusted R ²	Mean square residual
A	Attracted vehicles = $0.208 * \text{Students} + 19.12 \text{ Floor area}$	0.89	473
B	Attracted vehicles = $4.84 * \text{Employees}$	0.86	1166
C	Attracted vehicles = $36.47 * \text{Floor area}$	0.84	2112
D	Attracted vehicles = $0.40 * \text{Students}$	0.84	2188
E	Attracted vehicles = $13.72 * \text{Classes}$	0.83	5210
F	Attracted vehicles = $16.94 * \text{Lot area}$	0.77	2752

Table 2.46: Summary of the developed models for intermediary schools (Trip Attraction)

Model number	Developed models	Adjusted R^2	Mean square residual
A	Attracted vehicles = $0.56 * \text{Students}$	0.87	1120
B	Attracted vehicles = $57 * \text{Floor Area}$	0.87	1415
C	Attracted vehicles = $5.89 * \text{Employees}$	0.86	1671
D	Attracted vehicles = $19.72 * \text{Lot area}$	0.83	3624
E	Attracted vehicles = $15.57 * \text{Classes}$	0.82	4473

Table 2.47: Summary of the developed models for primary schools (trip attraction)

Model number	Developed model	Adjusted R^2	Mean square residual
A	Attracted vehicles = $0.34 * \text{Students} + 6.67 * \text{Lot area}$	0.86	509
B	Attracted vehicles = $5.4 * \text{Employees}$	0.86	1077
C	Attracted vehicles = $10.86 * \text{Classes}$	0.85	1420
D	Attracted vehicles = $0.465 * \text{Students}$	0.85	1568
E	Attracted vehicles = $61.29 * \text{Floor Area}$	0.84	1701
F	Attracted vehicles = $19.32 * \text{Lot area}$	0.72	6192

Table 2.48: Summary of the developed models for all levels of schools (Trip Attraction)

Model number	Developed model	Adjusted R^2	Mean square residual
A	Attracted vehicles = $0.252 \times \text{Students} + 20.51 \times \text{Floor area} + 3.34 \times \text{Lot area} - 61.79 \times \text{Dummy}$	0.94	908
B	Attracted vehicles = $5.29 \times \text{Employees}$	0.93	1526
C	Attracted vehicles = $0.316 \times \text{Students} + 6.83 \times \text{Lot area}$	0.93	1435
D	Attracted vehicles = $0.468 \times \text{Students}$	0.91	2355
E	Attracted vehicles = $13.16 \times \text{Classes}$	0.89	3629
F	Attracted vehicles = $46.44 \times \text{Floor area}$	0.88	3949
G	Attracted vehicles = $18.56 \times \text{Lot area}$	0.86	4860

Table 2.49: Summary of the developed models for primary & intermediary of schools (Trip Attraction)

Model number	Developed model	Adjusted R^2	Mean square residual
A	Attracted vehicles = $0.36 \times \text{Students} + 6.98 \times \text{Lot area}$	0.93	688
B	Attracted vehicles = $5.69 \times \text{Employees}$	0.92	1382
C	Attracted vehicles = $58.48 \times \text{Floor area}$	0.91	1530
D	Attracted vehicles = $0.52 \times \text{Students}$	0.91	1649
E	Attracted vehicles = $12.91 \times \text{Classes}$	0.86	4185
F	Attracted vehicles = $19.58 \times \text{Lot area}$	0.85	4655

By comparing the predicted values with the observed values for each level of schools, it is obvious that the worst models in predicting the number of attracted vehicles to schools are Riyadh city manual models since in most cases these models overestimated the number of attracted vehicles, especially when the variable used is the lot or floor area. The main reason for this overestimating is that the Riyadh city manual was prepared based on the data from schools of boys and girls. ITE and Dubai trip generation rates predicted the number of attracted vehicles much better than that predicted by Riyadh city rates. It is suggested to use the international models (ITE and Dubai trip generation rates) rather than the local models in the study area.

CHAPTER 3

MODELING OF PARKING GENERATION

3.1 Data Analysis and Discussion

Regression analysis is a statistical tool for the investigation of relationships between variables. Usually, the investigator seeks to ascertain the causal effect of one variable upon another.

In this study, single and multiple regressions were used to develop a model which predicts the parking demand for schools using many of the dependent variables

3.1.1 Experimental Design

A set of explanatory factors summarizing the characteristics of schools was evaluated. The relationship between these explanatory variables and the number of parked vehicles (dependent variable) was thoroughly investigated through regression. The explanatory variables are:

- The number of students
- The number of employees
- The number of classrooms
- Total floor area: the sum (in square meter) of the area of each floor level for all buildings.

- Total lot area: defined as the area (in square meter) of the built and unbuilt land owned by the school.
- Availability of parking space lots: was presented as a dummy variable in multiple regression analysis.

For each dependent variable, the analysis was done four times for each level of schools as described below:

1. Data analysis for seven schools with intercept
2. Data analysis for seven schools without intercept
3. Data analysis for ten schools with intercept
4. Data analysis for ten schools without intercept

The purpose of analyzing seven schools out of ten was to use the remaining three schools in the validation process, which can help in determining the best fit models. For each level of schools, the models which resulted from analyzing seven schools were compared with those that resulted from analyzing ten schools.

The same procedure was done when all schools were pooled together as described below:

1. Data analysis for 21 schools with intercept
2. Data analysis for 21 schools with intercept using two dummy variables to distinguish between the three levels of schools
3. Data analysis for 21 schools without intercept
4. Data analysis for 21 schools without intercept using two dummy variables to distinguish between the three levels of schools
5. Data analysis for 30 schools with intercept

6. Data analysis for 30 schools with intercept using two dummy variables to distinguish between the three levels of schools
7. Data analysis for 30 schools without intercept
8. Data analysis for 30 schools without intercept using two dummy variables to distinguish between the three levels of schools

The data were analyzed using the SPSS software.

3.1.2 Data Analysis for Secondary Schools

3.1.2.1 Summary of Results

The correlation between the dependent variable (parked vehicles) and independent variables for secondary schools was introduced by the following matrix (Table 3.1).

Table 3.1: Correlation matrix for the variables of secondary schools (10 schools)

	<i>Parked</i>	<i>Total Attracted</i>	<i>Students</i>	<i>Classes</i>	<i>Employees</i>	<i>Lot area</i>	<i>Floor area</i>	<i>Parking space area</i>	<i>Dist to nearest school</i>	<i>Dist to arterial</i>	<i>Dist to collector</i>	<i>Dist to CBD</i>
Parked	1.00											
Total Attracted	0.81	1.00										
Students	0.63	0.87	1.00									
Classes	0.41	0.76	0.94	1.00								
Employees	0.58	0.84	0.72	0.68	1.00							
lot area	0.31	0.44	0.42	0.36	0.50	1.00						
Floor area	0.42	0.72	0.43	0.42	0.66	0.58	1.00					
Parking space area	0.52	0.50	0.43	0.47	0.33	0.36	0.35	1.00				
Dist to nearest school	0.13	0.01	-0.26	-0.34	-0.40	-0.23	0.23	0.12	1.00			
Dist to arterial	0.04	0.14	0.02	-0.08	0.02	0.24	0.11	0.21	0.34	1.00		
Dist to collector	-0.02	-0.29	-0.57	-0.66	-0.59	-0.27	0.00	0.01	0.91	0.35	1.00	
Dist to CBD	0.12	-0.12	-0.30	-0.34	-0.57	-0.26	0.04	0.25	0.92	0.20	0.89	1.00

From the correlation matrix, it is clear that the most significant variable which has a strong linear relationship with the dependent variable (parked vehicles) is the number of students.

The relationships between the dependent variable y (parked vehicles) and independent variables for secondary schools are summarized in Tables 3.2 to 3.5 as shown below.

Table 3.2: Summary of the developed models with intercept for 7 secondary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.101*Students+28.69	0.51	506
Parked vehicles = 1.61*Employees-9.82	0.24	774
Parked vehicles = 3.06*Classes+32.79	0.25	768
Parked vehicles = 7.87*Floor area+34.03	0.03	996
Parked vehicles = 3.96*Lot area+29.62	0.10	919

Table 3.3: Summary of the developed models without intercept for 7 secondary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.149*Students	0.76	591
Parked vehicles = 1.83*Employees	0.75	652
Parked vehicles = 4.98*Classes	0.73	815
Parked vehicles = 13.485*Floor area	0.72	938
Parked vehicles = 6.18*Lot area	0.73	854

Table 3.4: Summary of the developed models with intercept for 10 secondary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.085*Students+42.62	0.32	558
Parked vehicles = 1.49*Employees+21	0.25	616
Parked vehicles = 2.08*Classes+53.15	0.06	772
Parked vehicles = 7.43*Floor area+42.06	0.08	762
Parked vehicles = 2.12*Lot area+58.56	-0.01	842

Table 3.5: Summary of the developed models without intercept for 10 secondary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.16*Students	0.79	830
Parked vehicles = 1.97*Employees	0.82	579
Parked vehicles = 5.48*Classes	0.86	1115
Parked vehicles = 14.76*Floor area	0.79	834
Parked vehicles = 6.93*Lot area	0.74	1200

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

The remaining three schools which were not included in developing the models of seven schools without intercept were used to validate these models. Table 3.6 includes the observed values of dependent variable in the first column and the predicted values by the developed models of seven schools without intercept in the other columns.

Table 3.6: Predicted values by different variables using the developed models (Secondary, Parking Generation)

Number of parked vehicles	Students (difference)	Employees (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)
Coefficient	0.15	1.83	4.98	13.48	6.18
114	67(-41%)	87(-24%)	50(-56%)	78(-32%)	57(-50%)
84	37(-56%)	55(-34%)	45(-46%)	65(-23%)	33(-61%)
70	62(-11%)	59(-16%)	65(-7%)	44(-37%)	47(-33%)
Average difference %	36	25	36	31	48
Range	11-56	16-34	7-56	23-37	33-61

CV: coefficient of variation = Standard deviation / Mean

$$\text{Difference} = \frac{\text{Predicted value} - \text{Observed value}}{\text{Observed value}} * 100\%$$

3.1.2.2 Discussion

By comparing the models which resulted from the seven schools analysis with those that resulted from the ten schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of ten schools analysis. On the other hand, there is no significant difference in the coefficients of independent variables

between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of parked vehicles when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 3.5 which summarized the results of analysis of ten schools without intercept and Table 3.6 which contains the predicted numbers of attracted vehicles using the developed models, the best model with the lowest mean square residual that predicts the number of parked vehicles at secondary schools is the following regression model:

$$\text{Model A: number of parked vehicles} = 1.97 * \text{Employees}$$

The model has the smallest mean square residual. The assumptions of normality, independency and equal variance of the model were checked graphically by considering the plot of residuals against the number of students, the plot of residuals against the total floor area, and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of parked vehicles were arranged in order (from the smallest mean square error to the largest one) as follows:

Model B: number of parked vehicles = 0.16 * Number of Students

Model C: number of parked vehicles = 14.76 * Total Floor Area

Model D: number of parked vehicles = 5.48 * Number of Classes

Model E: number of parked vehicles = 6.93 * Total Lot Area

The characteristics of these models are summarized in Table 3.7.

Table 3.7: Characteristics of the developed models (Secondary, Parking Generation)

Model number	Variable	Coefficient	P-Value	Mean square residual
A	Employees	1.97	1.79E-06	579
B	Students	0.16	9.17E-06	830
C	Floor Area	14.76	9.39E-06	834
D	Classes	5.48	3.53E-05	1115
E	Lot Area	6.93	4.93E-05	1200

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the independent variables and the normality plot of residuals. All assumptions seemed to be reasonable.

3.1.3 Data Analysis for Intermediary Schools

3.1.3.1 Summary of Results

The correlation between the dependent variable (parked vehicles) and independent variables for intermediary schools was introduced by the following matrix (Table 3.8).

Table 3.8: Correlation matrix for the variables of intermediary schools (10 schools)

	<i>Parked</i>	<i>Total Attracted</i>	<i>Students</i>	<i>Class- rooms</i>	<i>Employees</i>	<i>Lot area</i>	<i>Floor area</i>	<i>Parking space area</i>	<i>Dist to nearest school</i>	<i>Dist to arterial</i>	<i>Dist to collector</i>	<i>Dist to CBD</i>
Parked	1.00											
Total Attracted	0.84	1.00										
Students	0.86	0.94	1.00									
Classrooms	0.70	0.64	0.73	1.00								
Employees	0.84	0.88	0.91	0.65	1.00							
Lot area	0.62	0.83	0.76	0.25	0.85	1.00						
Floor area	0.82	0.91	0.89	0.64	0.97	0.83	1.00					
Parking space area	0.37	0.49	0.54	0.09	0.74	0.80	0.66	1.00				
Dist to nearest school	0.03	0.17	0.30	0.17	0.08	0.18	0.04	0.02	1.00			
Dist to arterial	-0.08	0.16	0.17	-0.16	0.05	0.17	0.02	0.32	-0.21	1.00		
Dist to collector	0.10	0.24	0.19	-0.27	-0.06	0.18	-0.01	-0.05	0.27	0.44	1.00	
Dist to CBD	-0.05	0.09	0.11	-0.19	-0.18	0.01	-0.21	-0.10	0.26	0.65	0.86	1.00

From the correlation matrix, it is clear that the most significant variables which have a strong linear relationship with the dependent variable (parked vehicles) are the number of students, employees and floor area.

The relationships between the dependent variable y (parked vehicles) and independent variables for intermediary schools are summarized in Tables 3.9 to 3.12 as shown below.

Table 3.9: Summary of the developed models with intercept for 7 intermediary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.0692*Students+7.7723	0.76	21.32
Parked vehicles = 0.7956*Employees+4.7799	0.83	15.47
Parked vehicles = 1.5464*Classes+12.985	0.31	61.84
Parked vehicles = 5.1654*Floor area+15.967	0.76	21.22
Parked vehicles = 1.1055*Lot area+25.235	0.39	54.48
Parked vehicles = 0.72 +1.01*Employees-6.56*Dummy	0.86	12.30

Where,

Dummy = 1, if parking lot areas are available, 0 otherwise

Table 3.10: Summary of the developed models without intercept for 7 intermediary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.0862*Students	0.82	22.42
Parked vehicles = 0.9057*Employees	0.82	14.45
Parked vehicles = 2.3295*Classes	0.80	61.83
Parked vehicles = 8.5721*Floor area	0.80	53.19
Parked vehicles = 2.9417*Lot area	0.72	199.05
Parked vehicles = 1.03*Employees-6.78*Dummy	0.86	12.30

Where,

Dummy = 1, if parking lot areas are available, 0 otherwise

Table 3.11: Summary of the developed models with intercept for 10 intermediary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.0683*Students+7.1746	0.70	22.87
Parked vehicles = 0.6772*Employees+8.9672	0.67	25.22
Parked vehicles = 1.3613*Classes+15.778	0.43	44.18
Parked vehicles = 4.7127*Floor area+16.749	0.64	28.00
Parked vehicles = 1.0482*Lot area+23.755	0.31	53.05
Parked vehicles = 3.65+1.001*Employees-9.964*Dummy	0.80	15.16

Where,

Dummy = 1, if parking lot areas are available, 0 otherwise

Table 3.12: Summary of the developed models without intercept for 10 intermediary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.085*Students	0.87	23.96
Parked vehicles = 0.90*Employees	0.87	28.62
Parked vehicles = 2.3973*Classes	0.84	63.43
Parked vehicles = 8.5479*Floor area	0.84	64.72
Parked vehicles = 2.9025*Lot area	0.77	163.07
Parked vehicles = 1.113*Employees- 11.128*Dummy	0.86	14.22

Where,

Dummy = 1, if parking lot areas are available, 0 otherwise

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

3.1.3.2 Models Validation

The remaining three schools which were not included in developing the models of seven schools without intercept were used to validate these models. Table 3.13 includes the observed values of dependent variable in the first column and the predicted values by the developed models of seven schools without intercept in the other columns.

Table 3.13: Predicted values by different variables using the developed models (Intermediary, Parking Generation)

Number of parked vehicles	Students (difference)	Employees (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)	Employees + Dummy (difference)
Coefficient	0.09	0.91	2.33	8.57	2.94	1.03,-6.78
30	24(-20%)	19(-37%)	17(-43%)	16(-47%)	16(-47%)	22(-26%)
28	34(+21%)	34(+21%)	23(-18%)	35(+25%)	33(+17%)	31(+11%)
28	32(+14%)	33(+18%)	28(0%)	28(0%)	35(+25%)	30(+7%)
Average difference %	18	25	20	24	30	15
Range %	14-21	18-37	0-43	0-47	17-47	7-26

CV: coefficient of variation = Standard deviation / Mean

$$\text{Difference} = \frac{\text{Predicted value} - \text{Observed value}}{\text{Observed value}} * 100\%$$

3.1.3.3 Discussion

By comparing the models which resulted from the seven schools analysis with those that resulted from the ten schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of ten schools analysis. On the other

hand, there is no significant difference in the coefficients of independent variables between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of attracted vehicular trips to school when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 3.12 which summarized the result of analysis of ten schools without intercept and Table 3.13 which contains the predicted numbers of attracted vehicles using the developed models, the best multiple and simple regression models with the lowest mean square residual that explains the attraction behavior to secondary schools are as follows:

Model A: number of attracted vehicles = $1.113 * \text{Employees} - 11.128 * \text{Dummy}$

Model B: number of parked vehicles = $0.085 * \text{Number of Students}$

The models have the smallest mean square residual. The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the number of classes, the plot of residuals against the total lot area, the plot of residuals against number employees, the plot of residuals against the

total floor area, and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of parked vehicles at intermediary schools were arranged in order as follows:

Model C: number of parked vehicles = 0.90 * Number of Employees

Model D: number of parked vehicles = 2.40 * Number of Classes

Model E: number of parked vehicles = 8.54 * Total Floor Area

Model F: number of parked vehicles = 2.90 * Total Lot Area

The characteristics of these models are summarized in Table 3.14 as shown below.

Table 3.14: Characteristics of the developed models (Intermediary, Parking Generation)

Model number	Variable	Coefficient	P-Value
A	Employees	1.113	3.9E-07
	Dummy	-11.128	1.0E-02
B	Students	0.085	2.78E-09
C	Employees	0.90	6.21E-09
D	Classes	2.40	2.25E-07
E	Floor Area	8.54	2.47E-07
F	Lot Area	2.90	1.63E-05

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the number of students and the normality plot of residuals. All assumptions seemed to be reasonable.

3.1.4 Data Analysis for Primary Schools (Parking Generation)

3.1.4.1 Summary of Results

The correlation between the dependent variable (parked vehicles) and independent variables for primary schools was introduced by the following matrix (Table 3.15).

Table 3.15: Correlation matrix for the variables of primary schools (10 schools)

	<i>Parked</i>	<i>Total Attracted</i>	<i>Students</i>	<i>Classes</i>	<i>Employees</i>	<i>Lot area</i>	<i>Floor area</i>	<i>Parking space area</i>	<i>Dist to nearest school</i>	<i>Dist to arterial</i>	<i>Dist to collector</i>	<i>Dist to CBD</i>
Parked	1.00											
Total Attracted	0.87	1.00										
Students	0.91	0.85	1.00									
Classes	0.96	0.86	0.97	1.00								
Employees	1.00	0.88	0.92	0.97	1.00							
Lot area	0.62	0.78	0.42	0.52	0.63	1.00						
Floor area	0.84	0.87	0.78	0.83	0.86	0.80	1.00					
Parking space area	0.66	0.66	0.48	0.49	0.64	0.74	0.67	1.00				
Dist to nearest school	-0.57	-0.46	-0.57	-0.56	-0.56	-0.26	-0.62	-0.45	1.00			
Dist to arterial	0.17	0.36	0.04	0.08	0.17	0.47	0.38	0.41	-0.05	1.00		
Dist to collector	-0.17	-0.02	-0.16	-0.17	-0.16	-0.02	-0.11	-0.06	0.58	0.65	1.00	
Dist to CBD	-0.03	0.14	-0.11	-0.05	-0.01	0.33	0.01	0.03	0.70	0.50	0.74	1.00

From the correlation matrix, it is clear that the most significant variables which have a strong linear relationship with the dependent variable (parked vehicles) are the number of employees.

The relationships between the dependent variable y (parked vehicles) and independent variables for primary schools are summarized in Tables 3.16 to 3.19 as shown below.

Table 3.16: Summary of the developed models with intercept for 7 primary schools (Parking Generation)

Model	Adjusted R^2	Mean square residual
Parked vehicles = $0.072 \times \text{Students} + 4.568$	0.81	35.13
Parked vehicles = $0.9714 \times \text{Employees} - 0.7985$	1.00	0.89
Parked vehicles = $1.8698 \times \text{Classes} + 0.903$	0.95	8.72
Parked vehicles = $8.2884 \times \text{Floor area} + 9.4049$	0.68	60.23
Parked vehicles = $1.3258 \times \text{Lot area} + 20.762$	0.26	139.21
Parked vehicles = $-1.28 - 0.057 \times \text{Classes} + 1.296 \times \text{Employees} - 0.128 \times \text{Lot area}$	1.00	0.73

Table 3.17: Summary of the developed models without intercept for 7 primary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.0826*Students	0.81	33.42
Parked vehicles = 0.95*Employees	0.83	0.85
Parked vehicles = 1.9185*Classes	0.83	7.41
Parked vehicles = 11.228*Floor area	0.78	72.14
Parked vehicles = 3.05*Lot area	0.58	318.90

Table 3.18: Summary of the developed models with intercept for 10 primary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.0678*Students+5.5205	0.78	28.14
Parked vehicles = 0.9587*Employees-0.406	0.99	0.97
Parked vehicles = 1.7082*Classes+3.4646	0.91	12.56
Parked vehicles = 7.5405*Floor area+9.6892	0.66	46.77
Parked vehicles = 1.3483*Lot area+20.256	0.314	95.32
Parked vehicles = -1.37- 0.625*Classes+1.32*Employees-0.130*Lot area	0.99	0.38

Table 3.19: Summary of the developed models without intercept for 10 primary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.081*Students	0.86	30.10
Parked vehicles = 0.9473*Employees	0.89	0.89
Parked vehicles = 1.902*Classes	0.88	13.08
Parked vehicles = 10.546*Floor area	0.83	60.46
Parked vehicles = 3.2233*Lot area	0.66	248.16

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

3.1.4.2 Models Validation

The remaining three schools which were not included in developing the models of seven schools without intercept were used to validate these models. Table 3.20 includes the observed values of dependent variable in the first column and the predicted values by the developed models of seven schools without intercept in the other columns.

Table 3.20: Predicted values by different variables using the developed models (Primary, Parking Generation)

Number of parked vehicles	Students (difference)	Employees (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)
Coefficient	0.08	0.95	1.92	11.23	3.05
29	36(+24%)	30(+3%)	34(+17%)	35(+21%)	12(-59%)
20	16(-20%)	19(-5%)	13(-35%)	18(-10%)	16(-20%)
35	35(0%)	35(0%)	35(0%)	45(+29%)	31(-11%)
Average difference %	15	3	17	20	30
Range	0-24	0-5	0-35	10-29	11-59

CV: coefficient of variation = Standard deviation / Mean

$$\text{Difference} = \frac{\text{Predicted value} - \text{Observed value}}{\text{Observed value}} * 100\%$$

3.1.4.3 Discussion

By comparing the models which resulted from the seven schools analysis with those that resulted from the ten schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of ten schools analysis. On the other hand, there is no significant difference in the coefficients of independent variables

between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of parked vehicles when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 3.19 which summarized the results of analysis of ten schools without intercept and Table 3.20 which contains the predicted numbers of attracted vehicles using the developed models, the best model with the lowest mean square residual that predicts the number of parked vehicles at primary schools is the following regression model:

$$\text{Model A: number of parked vehicles} = 0.95 * \text{Number of Employees}$$

The model has the smallest mean square residual. The assumptions of normality, independency and equal variance of the model were checked graphically by considering the plot of residuals against the number of students, the plot of residuals against the total floor area, and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of parked vehicles were arranged in order (from the lowest mean square error to the largest one) as follows:

Model B: number of parked vehicles = 1.90 * Number of Classes

Model C: number of parked vehicles = 0.081 * Number of students

Model D: number of parked vehicles = 10.55 * Total Floor Area

Model E: number of parked vehicles = 3.22 * Total Lot Area

The characteristics of these models are summarized in Table 3.21.

Table 3.21: Characteristics of the developed models (Primary, Parking Generation)

Model number	Variable	Coefficient	P-Value	Mean square residual
A	Employees	0.95	2.96E-15	0.89
B	Classes	1.90	5.39E-10	13.08
C	Students	0.08	2.31E-08	30.10
D	Floor Area	10.55	5.38E-07	60.46
E	Lot Area	3.22	3.0E-04	248.16

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the independent variables and the normality plot of residuals. All assumptions seemed to be reasonable.

3.1.5 Data Analysis for All Levels Schools (Parking Generation)

3.1.5.1 Summary of Results

The simple regression analysis for all levels schools was done in the same procedure as each level of schools. The only difference is in multiple regression analysis which was done two times. The first time was done without distinguishing between the levels of schools. In the second multiple regression analysis, two dummy variables were used to distinguish between the levels of schools as follows:

Dummy1 = 1 if secondary school, 0 otherwise

Dummy2 = 1 if intermediary school, 0 otherwise

The correlation between the dependent variable (parked vehicles) and independent variables for all schools pooled together was introduced by the following matrix (Table 3.22).

Table 3.22: Correlation matrix for the variables of all levels pooled together (30 schools)

	<i>Parked</i>	<i>Total Attracted</i>	<i>Students</i>	<i>Classes</i>	<i>Employees</i>	<i>Lot area</i>	<i>Floor area</i>	<i>Parking space area</i>	<i>Dist to nearest school</i>	<i>Dist to arterial</i>	<i>Dist to collector</i>	<i>Dist to CBD</i>
Parked	1.00											
Total Attracted	0.42	1.00										
Students	0.58	0.80	1.00									
Classes	0.24	0.66	0.82	1.00								
Employees	0.57	0.84	0.81	0.69	1.00							
Lot area	0.39	0.72	0.51	0.31	0.70	1.00						
Floor area	0.70	0.71	0.63	0.40	0.80	0.72	1.00					
Parking space area	0.46	0.53	0.47	0.30	0.62	0.70	0.60	1.00				
Dist to nearest school	0.25	-0.06	-0.19	-0.34	-0.22	-0.06	0.16	0.00	1.00			
Dist to arterial	0.03	0.24	0.07	-0.04	0.10	0.32	0.16	0.31	0.13	1.00		
Dist to collector	-0.07	-0.05	-0.26	-0.30	-0.27	-0.07	-0.09	-0.06	0.55	0.48	1.00	
Dist to CBD	-0.04	0.04	-0.15	-0.17	-0.24	0.04	-0.09	0.01	0.61	0.43	0.77	1.00

From the correlation matrix, it is clear that the most significant variable which has a strong linear relationship with the dependent variable (parked vehicles) is floor area.

The relationships between the dependent variable y (parked vehicles) and independent variables for all levels schools are summarized in Tables 3.23 to 3.26 as shown below.

Table 3.23: Summary of the developed models with intercept for 21 pooled schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.1113*Students+1.27	0.47	447
Parked vehicles = 1.41*Employees-5.58	0.34	560
Parked vehicles = 1.94*Classes+18.60	0.10	766
Parked vehicles = 10.48*Floor area+5.88	0.49	438
Parked vehicles = 2.58*Lot area+22.42	0.22	667
Parked vehicles = 0.15*Students+4.84*Floor area-2.95*Classes+8.63	0.66	288
Parked vehicles = 0.088*Students+35.60*Dummy1-0.78	0.81	164

where,

Dummy1 = 1 if secondary, 0 otherwise

Table 3.24: Summary of the developed models without intercept for 21 pooled schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.114*Students	0.82	425
Parked vehicles = 1.278*Employees	0.79	535
Parked vehicles = 3.01*Classes	0.71	771
Parked vehicles = 11.63*Floor area	0.82	423
Parked vehicles = 4.29*Lot area	0.72	757
Parked vehicles = 0.15*Students+5.57*Floor area-2.45*Classes	0.86	280
Parked vehicles = 0.087*Students+35.57*Dummy1	0.90	155

where,

Dummy1 = 1 if secondary, 0 otherwise

Table 3.25: Summary of the developed models with intercept for 30 pooled schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.105*Students+6.09	0.32	604
Parked vehicles = 1.44*Employees-4.33	0.30	616
Parked vehicles = 1.277*Classes+30.36	0.02	863
Parked vehicles = 11.47*Floor area+3.67	0.49	464
Parked vehicles = 2.25*Lot area+27.47	0.12	777
Parked vehicles = 0.15*Students-3.26*Classes-1.83*Lot area+10.63*Floor area+9.96	0.62	332
Parked vehicles = 0.08*Students+43.4*Dummy1+2.58	0.79	182

where,

Dummy1 = 1 if secondary, 0 otherwise

Table 3.26: Summary of the developed models without intercept for 30 pooled schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.118*Students	0.79	588
Parked vehicles = 1.337*Employees	0.79	596
Parked vehicles = 3.14*Classes	0.68	954
Parked vehicles = 12.25*Floor area	0.83	450
Parked vehicles = 4.52*Lot area	0.67	925
Parked vehicles = 0.15*Students-2.82*Classes- 1.73*Lot area+11.29*Floor area	0.86	331
Parked vehicles = 0.083*Students+43.5*Dummy1	0.91	177

where,

Dummy1 = 1 if secondary, 0 otherwise

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

3.1.5.2 Models Validation

The remaining nine schools which were not included in developing the models were used to validate these models. Table 3.27 includes the observed values of dependent variable in the first column and the predicted values by all the developed models in the other columns.

Table 3.27: Predicted values by different variables using the developed models (All Levels, Parking Generation)

Number of parked vehicles	Employees (difference)	Students (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)	Classes + Students + Floor Area (difference)	Students + Dummy1 (difference)
Coefficient	1.28	0.11	3.01	11.63	4.29	0.15,5.57,2.45	0.09,35.57
114	61(-46%)	51(-55%)	30(-73%)	67(-41%)	40(-65%)	75(-34%)	75(-34%)
84	38(-55%)	29(-65%)	27(-68%)	56(-33%)	23(-73%)	42(-50%)	57(-32%)
70	40(-43%)	47(-33%)	39(-44%)	38(-45%)	33(-53%)	48(-31%)	71(+1%)
30	27(-10%)	32(+7%)	21(-30%)	21(-30%)	23(-23%)	35(+17%)	24(-20%)
28	47(+68%)	44(+57%)	30(+7%)	48(+71%)	48(+71%)	57(+100%)	34(+21%)
28	46(+64%)	43(+54%)	36(+29%)	38(+36%)	50(+79%)	45(+61%)	33(+18%)
29	41(+41%)	50(+72%)	54(+86%)	36(+24%)	16(-45%)	39(+34%)	38(+31%)
20	26(+30%)	21(+5%)	21(+5%)	19(-5%)	22(+10%)	20(0%)	16(-20%)
35	47(+34%)	48(+37%)	54(+54%)	47(+34%)	44(+25%)	41(+17%)	36(+3%)
Average difference %	44	43	44	36	50	39	20
Range	10-68	5-72	5-86	5-71	10-79	0-100	1-34

3.1.5.3 Discussion

By comparing the models which resulted from the 21 schools analysis with those that resulted from the 30 schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of 30 schools analysis. On the other hand, there is no significant difference in the coefficients of independent variables between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of attracted vehicular trips to school when one of the independent variables is zero. The statistical measure of goodness of fit of the regression models that was used in this study is the mean square residual which measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 3.26 which summarized the results of analysis of 30 schools without intercept and Table 3.27 which contains the predicted numbers of attracted vehicles using the developed models, the best multiple and simple regression models with the lowest mean square residual that explains the attraction behavior to secondary schools are as follows:

Model A: number of parked vehicles = $0.083 * \text{Students} + 43.5 * \text{Dummy1}$

where, Dummy1 = 1 if secondary, 0 otherwise

Model B: number of parked vehicles = 12.25 * Total Floor Area

The model has the highly significant coefficients with P-values of 1.72E-12 for the coefficient of the number of students and 5.46E-9 for the dummy variable. The assumptions of normality, independency and equal variance of the model were checked graphically by considering the plot of residuals against the number of employees and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of parked vehicles at schools were arranged in order (from the smallest mean square residual to the largest one) as follows:

**Model C: number of parked vehicles = 0.15*Students-2.82*Classes-
1.73*Lot Area+11.29*Floor Area**

Model D: number of parked vehicles = 0.118 * Number of Students

Model E: number of parked vehicles = 1.337 * Number of Employees

Model F: number of parked vehicles = 4.52 * Total Lot Area

Model G: number of parked vehicles = 3.14 * Number of Classes

The characteristics of these models are summarized in Table 3.28 as shown below.

Table 3.28: Characteristics of the developed models (All Levels, Parking Generation)

Model number	Variable	Coefficient	P-Value	Mean square residual
A	Students	0.083	1.72E-12	177
	Dummy 1	43.50	5.46E-9	
B	Floor Area	12.25	4.19E-14	450
C	Students	0.15	0.0023	331
	Lot Area	1.73	0.0829	
	Classes	-2.82	0.0096	
	Floor Area	11.29	0.0007	
D	Students	0.118	2.03E-12	588
E	Employees	1.337	2.48E-12	596
F	Lot Area	4.52	1.55E-9	925
G	Classes	3.14	2.44E-09	954

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the independent variables and the normality plot of residuals. All assumptions seemed to be reasonable.

From **model A**, it is clear that the secondary schools have the significant effect because of the appearance of the dummy variable which is related to secondary schools. Because of that, another trial of analysis in which the primary and intermediary schools were pooled together, was performed as shown in the next section.

3.1.6 Data Analysis for Primary & Intermediary Schools

3.1.6.1 Summary of Results

The correlation between the dependent variable (parked vehicles) and independent variables for primary and intermediary schools was introduced by the following matrix (Table 3.29).

Table 3.29: Correlation matrix for the variables of primary & intermediary schools together (20 schools)

	<i>Parked</i>	<i>Total Attracted</i>	<i>Students</i>	<i>Class- rooms</i>	<i>Employees</i>	<i>Lot area</i>	<i>Floor area</i>	<i>Parking space area</i>	<i>Dist to nearest school</i>	<i>Dist to arterial</i>	<i>Dist to collector</i>	<i>Dist to CBD</i>
Parked	1.00											
Total Attracted	0.84	1.00										
Students	0.89	0.85	1.00									
Classrooms	0.80	0.64	0.85	1.00								
Employees	0.94	0.88	0.91	0.76	1.00							
Lot area	0.65	0.82	0.57	0.34	0.76	1.00						
Floor area	0.82	0.90	0.80	0.60	0.91	0.83	1.00					
Parking space area	0.55	0.56	0.51	0.31	0.68	0.76	0.65	1.00				
Dist to nearest school	-0.36	-0.15	-0.31	-0.37	-0.30	-0.07	-0.28	-0.27	1.00			
Dist to arterial	0.11	0.29	0.11	-0.03	0.16	0.36	0.24	0.38	-0.08	1.00		
Dist to collector	-0.13	0.02	-0.10	-0.17	-0.16	-0.01	-0.12	-0.07	0.50	0.54	1.00	
Dist to CBD	-0.04	0.10	-0.03	-0.10	-0.09	0.16	-0.10	-0.04	0.53	0.55	0.74	1.00

From the correlation matrix, it is clear that the most significant variables which have a strong linear relationship with the dependent variable (parked vehicles) are the number of employees, students and floor area.

The relationships between the dependent variable y (parked vehicles) and independent variables for primary and intermediary schools pooled together are summarized in Tables 3.30 to 3.33 as shown below.

Table 3.30: Summary of the developed models with intercept for 14 schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.072*Students+5.25	0.82	25
Parked vehicles = 0.89*Employees+1.35	0.94	8
Parked vehicles = 1.79*Classes+5.60	0.68	44
Parked vehicles = 5.97*Floor area+13.93	0.70	42
Parked vehicles = 1.28*Lot area+22.17	0.41	83
Parked vehicles = 0.73*Employees+0.48*Classes-0.38	0.96	5

Table 3.31: Summary of the developed models without intercept for 14 schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.085*Students	0.90	27
Parked vehicles = 0.93*Employees	0.92	8
Parked vehicles = 2.11*Classes	0.89	44
Parked vehicles = 9.35*Floor area	0.87	79
Parked vehicles = 2.98*Lot area	0.75	239
Parked vehicles = 0.73*Employees+0.47*Classes	0.91	5

Table 3.32: Summary of the developed models with intercept for 20 schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.07*Students+5.90	0.78	23
Parked vehicles = 0.83*Employees+3.47	0.87	14
Parked vehicles = 1.51*Classes+10.10	0.63	41
Parked vehicles = 5.66*Floor area+13.90	0.65	38
Parked vehicles = 1.23*Lot area+21.46	0.40	67
Parked vehicles = 0.68*Employees+0.40*Classes+2.61	0.88	13

Table 3.33: Summary of the developed models without intercept for 20 schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.083*Students	0.93	26
Parked vehicles = 0.92*Employees	0.93	15
Parked vehicles = 2.12*Classes	0.90	51
Parked vehicles = 9.22*Floor area	0.89	71
Parked vehicles = 3.02*Lot area	0.78	197
Parked vehicles = 0.73*Employees+0.45*Classes	0.93	13

When the results of analysis using intercept and the results of analysis without intercept were compared, it was obvious that the coefficients of independent variables were not significantly different. The results of ten schools without intercept were used as the final results to make it easy to compare with the international models which use ratios in most situations.

3.1.6.2 Models Validation

The remaining three schools which were not included in developing the models of seven schools without intercept were used to validate these models. Table 3.34 includes the observed values of dependent variable in the first column and the predicted values by the developed models of seven schools without intercept in the other columns.

Table 3.34: Predicted values by different variables using the developed models (Primary & Intermediary, Parking Generation)

Number of parked vehicles	Students (difference)	Employees (difference)	Classes (difference)	Floor Area (difference)	Lot Area (difference)	Employees + Classes (difference)
Coefficient	0.085	0.93	2.11	9.35	2.98	0.73,0.47
30	24(-20%)	20(-33%)	15(-50%)	17(-43%)	16(-47%)	19(-37%)
28	33(+18%)	34(+21%)	21(-25%)	38(+36%)	33(+18%)	32(+14%)
28	32(+14%)	33(+18%)	25(-10%)	31(+11%)	35(+25%)	32(+14%)
29	37(+28%)	30(+3%)	38(+31%)	29(0%)	11(-62%)	32(+10%)
20	16(-20%)	19(-5%)	15(-25%)	15(-25%)	15(-25%)	18(-10%)
35	36(+3%)	34(-3%)	38(+9%)	38(+9%)	30(-14%)	35(0%)
Average difference %	17	14	25	21	32	14
Range%	2-28	3-33	9-50	0-43	14-62	0-37

3.1.6.3 Discussion

By comparing the models which resulted from the 14 schools analysis with those that resulted from the 20 schools analysis, it can be concluded that the study models are stable since there is a small difference between the coefficients of independent variables. Based on that, the study depended on the results of 20 schools analysis. On the other hand, there is no significant difference in the coefficients of independent variables between the models built using the intercept and those in which the intercept was forced to be zero. This study depended on the models in which the intercept was forced to be zero to make it easier to compare the results with the local and international models. Furthermore, it is impractical to have a number of attracted vehicular trips to school when one of the independent variables is zero. The statistical measure of goodness of fit of the regression

models that was used in this study is the mean square residual which is measures the average of the squares of the “errors” [28]. The mean square error was used as a measure of goodness of fit rather than the coefficient of determination (R^2) to make it possible to compare between the models with intercept and those models without intercept.

Based on Table 3.33 which summarized the results of analysis of 20 schools without intercept and Table 3.34 which contains the predicted numbers of attracted vehicles using the developed models, the best multiple and simple regression models with the lowest mean square residual that explains the attraction behavior to secondary schools are as follows:

Model A: number of parked vehicles = $0.73 * \text{Employees} + 0.45 * \text{Classes}$

Model B: number of parked vehicles = $0.92 * \text{Number of Employees}$

The model has the highly significant coefficients with P-values of 4.75E-07 for the number of employees and 6.2E-03 for the number of classes. The model has the smallest mean square residual and largest F-value. The assumptions of normality, independency and equal variance of the model were checked graphically by considering the plot of residuals against the number of students, the plot of residuals against the total lot area, and the normality plot of residuals. All assumptions seemed to be reasonable.

The other models which are significant in predicting the number of parked vehicles at primary and intermediary schools were arranged in order (from the smallest mean square error to the largest one) as follows:

Model C: number of parked vehicles = $0.083 * \text{Number of Students}$

Model D: number of parked vehicles = 2.12 * Number of Classes

Model E: number of parked vehicles = 9.22 * Total Floor Area

Model F: number of parked vehicles = 3.02 * Total Lot Area

The characteristics of these models are summarized in Table 3.35 as shown below.

Table 3.35: Characteristics of the developed models (Primary & Intermediary, Parking Generation)

Model number	Variable	Coefficient	P-Value	Mean square residual
A	Employees	0.73	4.75E-07	13
	Classes	0.45	6.2E-03	
B	Employees	0.92	1.41E-19	15
C	Students	0.083	3.17E-17	26
D	Classes	2.12	1.980E-14	51
E	Floor Area	9.22	4.17E-13	71
F	Lot Area	3.02	7.150E-09	197

The assumptions of normality, independency and equal variance of the models were checked graphically by considering the plot of residuals against the independent variables and the normality plot of residuals. All assumptions seemed to be reasonable.

3.2 Summary of Results

Tables 3.36 to 3.40 represent the best models developed by the study for each type of analysis with their coefficient of determination and mean square residual.

Table 3.36: Summary of the developed models for 10 secondary schools (parking generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 1.97*Employees	0.82	579
Parked vehicles = 0.16*Students	0.79	830
Parked vehicles = 14.76*Floor area	0.79	834
Parked vehicles = 5.48*Classes	0.86	1115
Parked vehicles = 6.93*Lot area	0.74	1200

Table 3.37: Summary of the developed models for 10 intermediary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.085*Students	0.87	23.96
Parked vehicles = 0.90* Employees	0.87	28.62
Parked vehicles = 2.3973*Classes	0.84	63.43
Parked vehicles = 8.5479*Floor area	0.84	64.72
Parked vehicles = 2.9025*Lot area	0.77	163.07
Parked vehicles = 1.113*Employees- 11.128*Dummy	0.86	14.22

Table 3.38: Summary of the developed models for 10 primary schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.081*Students	0.86	30.10
Parked vehicles = 0.9473*Employees	0.89	0.89
Parked vehicles = 1.902*Classes	0.88	13.08
Parked vehicles = 10.546*Floor area	0.83	60.46
Parked vehicles = 3.2233*Lot area	0.66	248.16

Table 3.39: Summary of the developed models for 30 pooled schools (Parking Generation)

Model	Adjusted R²	Mean square residual
Parked vehicles = 0.118*Students	0.79	588
Parked vehicles = 1.337*Employees	0.79	596
Parked vehicles = 3.14*Classes	0.68	954
Parked vehicles = 12.25*Floor area	0.83	450
Parked vehicles = 4.52*Lot area	0.67	925
Parked vehicles = 0.15*Students- 2.82*Classes-1.73*Lot area+11.29*Floor area	0.86	331
Parked vehicles = 0.083*Students+43.5*Dummy1	0.91	177

Table 3.40: Summary of the developed models without intercept for 20 schools (Parking Generation)

Model	Adjusted R^2	Mean square residual
Parked vehicles = $0.083 \times \text{Students}$	0.93	26
Parked vehicles = $0.92 \times \text{Employees}$	0.93	15
Parked vehicles = $2.12 \times \text{Classes}$	0.90	51
Parked vehicles = $9.22 \times \text{Floor area}$	0.89	71
Parked vehicles = $3.02 \times \text{Lot area}$	0.78	197
Parked vehicles = $0.73 \times \text{Employees} + 0.45 \times \text{Classes}$	0.93	13

CHAPTER 4

QUESTIONNAIRE SURVEY ANALYSIS

4.1 Introduction

A questionnaire survey was conducted with the students for secondary schools and with the students' parents for intermediary and primary schools to study the socioeconomic characteristics of the students and to evaluate the appropriateness of some remedies in alleviating school congestion in the study area.

4.2 Procedure

The questionnaire was distributed over the 30 schools of the study. One hundred questionnaires were distributed for each school with a total of 3000 questionnaires. Tables 4.1 to 4.4 show the number of distributed, returned, invalid and analyzed questionnaires for each level of schools.

Table 4.1: Questionnaire statistics for secondary schools

Distributed questionnaires	1000
Returned questionnaires	757
Invalid questionnaires	24
Analyzed questionnaires	733
% Analyzed	73%

Table 4.2: Questionnaire statistics for intermediary schools

Distributed questionnaires	1000
Returned questionnaires	532
Invalid questionnaires	22
Analyzed questionnaires	510
% Analyzed	51%

Table 4.3: Questionnaire statistics for primary schools

Distributed questionnaires	1000
Returned questionnaires	572
Invalid questionnaires	12
Analyzed questionnaires	560
% Analyzed	56%

Table 4.4: Questionnaire statistics for all levels

Distributed questionnaires	3000
Returned questionnaires	1861
Invalid questionnaires	58
Analyzed questionnaires	1803
% Analyzed	60%

The invalid questionnaires are those which have some conflicts in the answers. (For example, when the student answered the question about the distance from home to school with 5-7 km and then answered the question about the time to arrive to school with 60 min, it means that he needs 60 min to pass a distance of 5-7 km by car, which is not reasonable.) It is obvious that the percentage of analyzed questionnaires for secondary schools is higher than those for primary and intermediary since the secondary school students filled the questionnaire by themselves, but for primary and intermediary, it was filled by the parents.

4.3 Descriptive Analysis

A descriptive analysis is a conclusion technique that is used to describe some key features of data in a research study. It gives a simple summary about a sample from the study. The following charts (Figures 4.1 to 4.56) present the results of analysis for each level of schools.

4.3.1 Secondary Schools

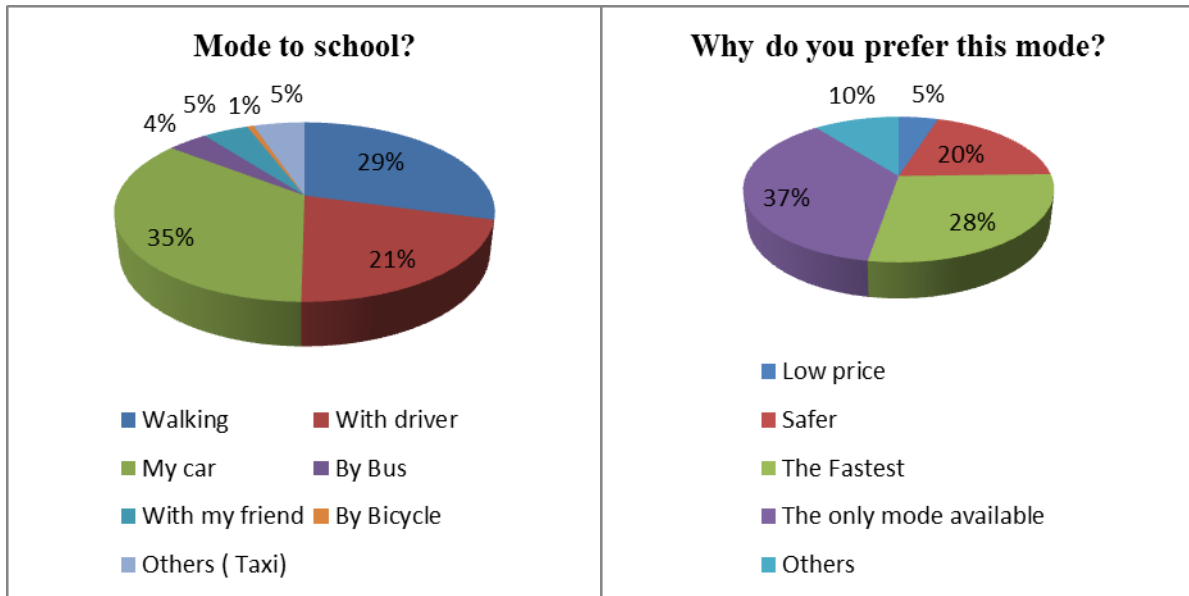


Figure 4.1: Mode to school

Figure 4.2: The reason for choosing this mode

From Figure 4.1, it is clear that 56% of the students of secondary schools are coming to the schools by passenger cars.

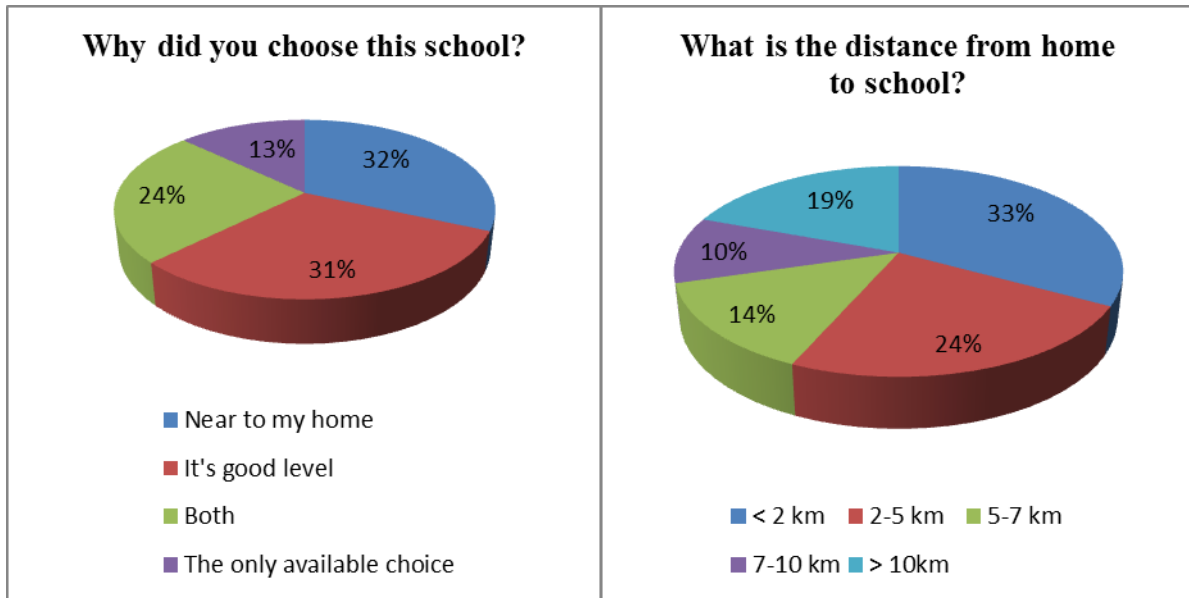


Figure 4.3: The reason for choosing this school

Figure 4.4: Distance from home to school

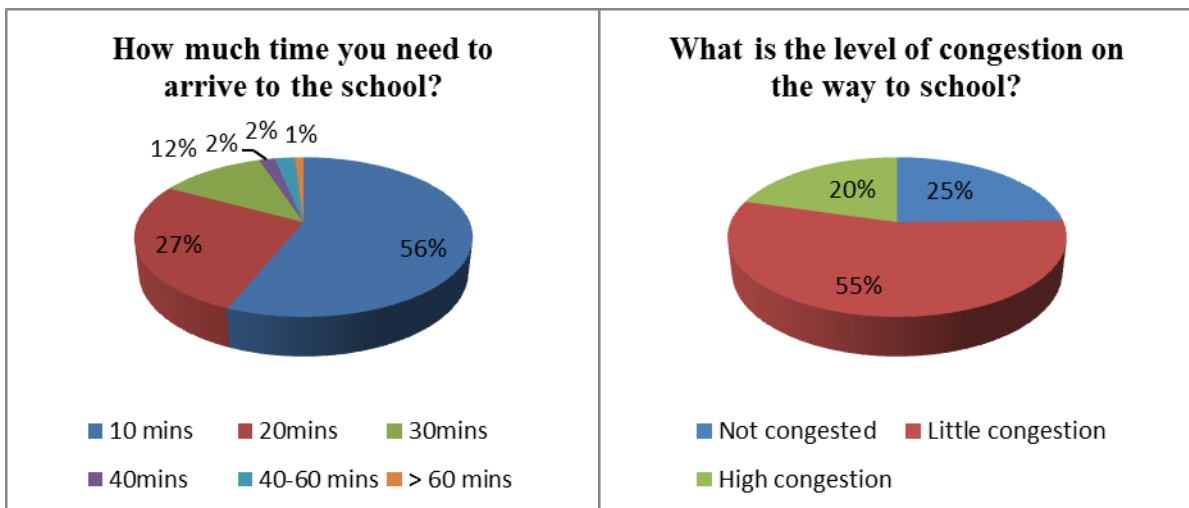


Figure 4.5: Time required to arrive to the school

Figure 4.6: Level of congestion on the way to school

From Figures 4.4 and 4.5, the average distance from home to school is 4.89 km, and the average time required to arrive to the school is 17 min.

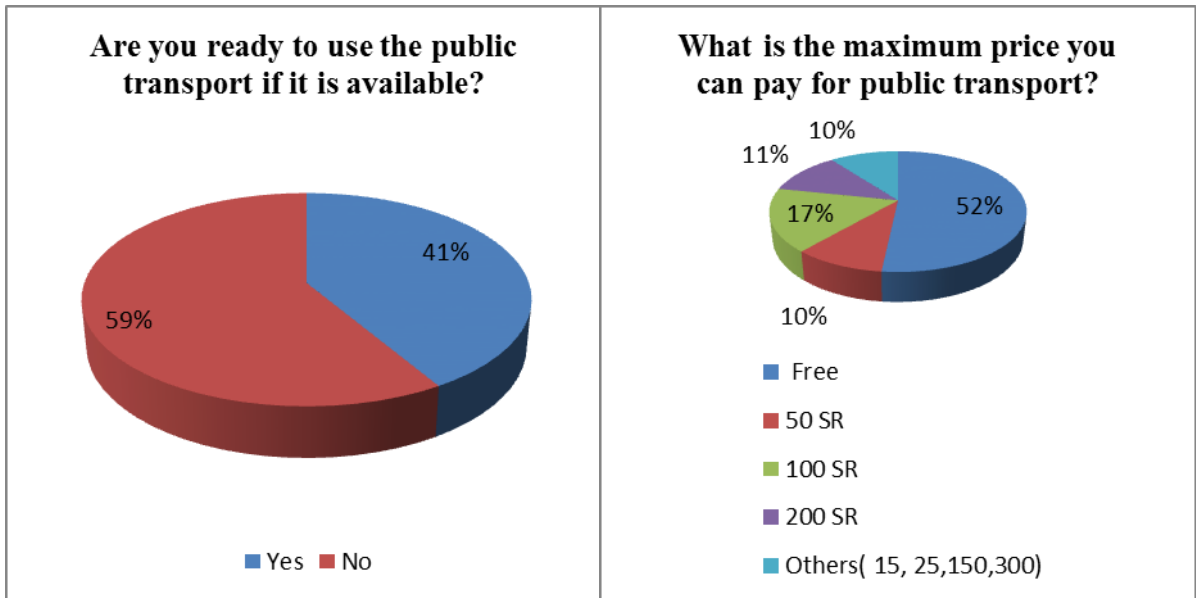


Figure 4.7: Willingness to use public transport

Figure 4.8: The maximum price for public transport

From Figures 4.7 and 4.8, 41% of students of secondary schools are ready to use public transportation, and 52% of them prefer this service to be free.

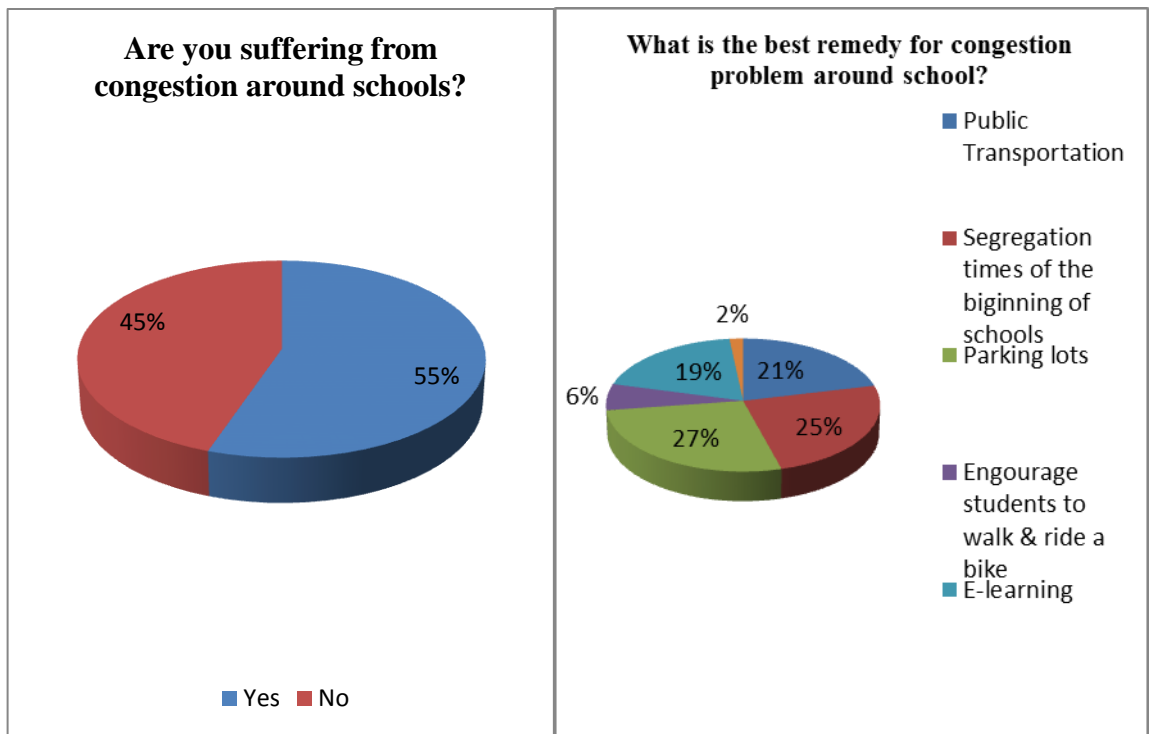


Figure 4.9: Suffering from congestion

Figure 4.10: Best remedies to solve congestion problem

According to the students of secondary schools, the best remedies to solve the congestion problem are providing the schools with more parking lots, segregation times of the beginning of schools, and public transportation, as shown in Figure 4.10.

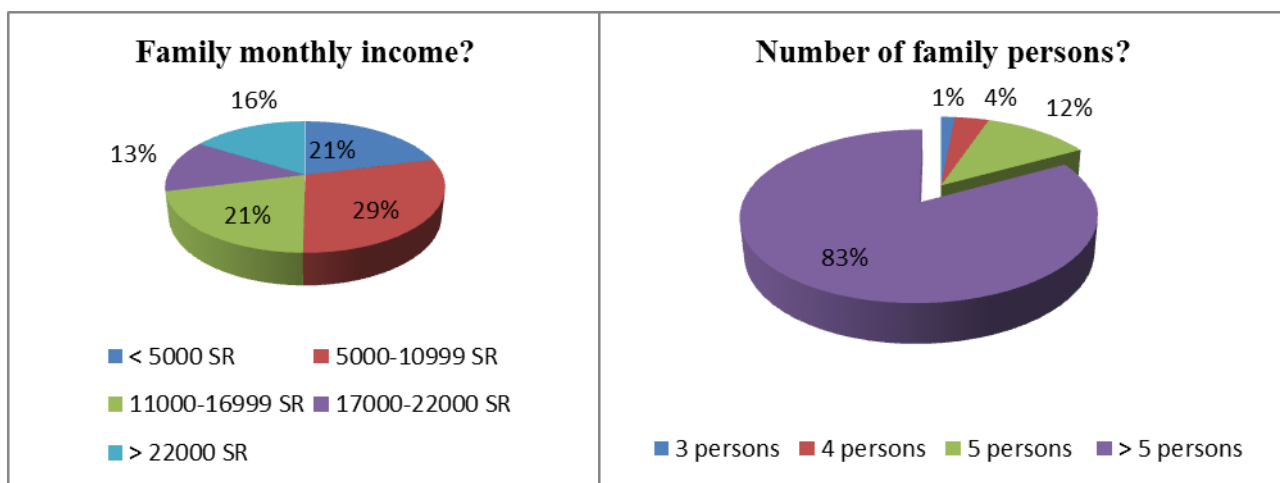


Figure 4.11: Family monthly income

Figure 4.12: Number of family persons

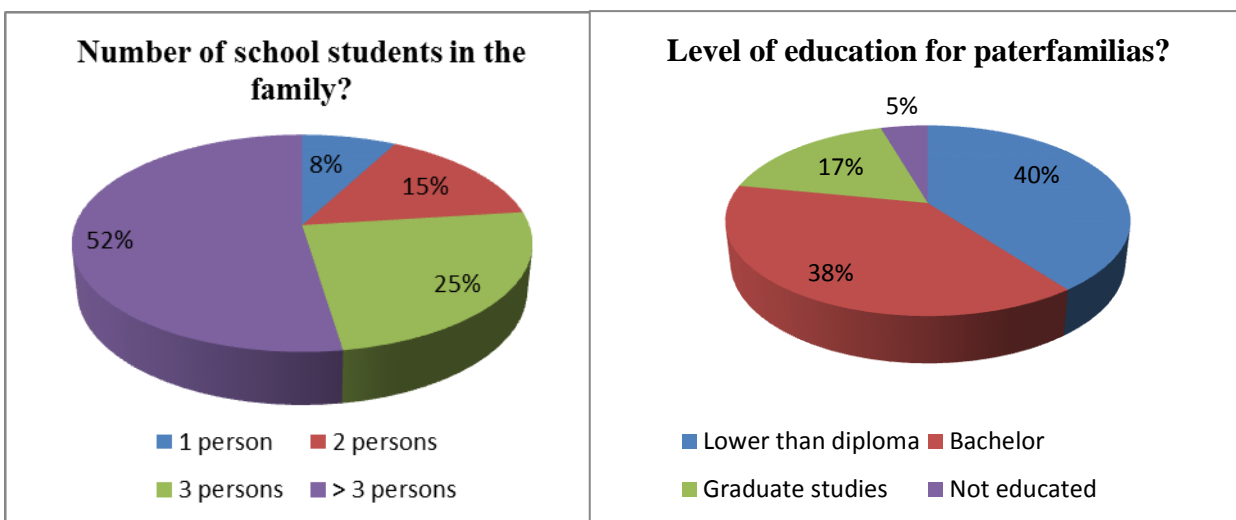


Figure 4.13: Number of students in the family

Figure 4.14: Level of education for paterfamilias

The weighted average for the family monthly income for students of secondary schools is 12400 SR as shown in Figure 4.11.

4.3.2 Intermediary Schools

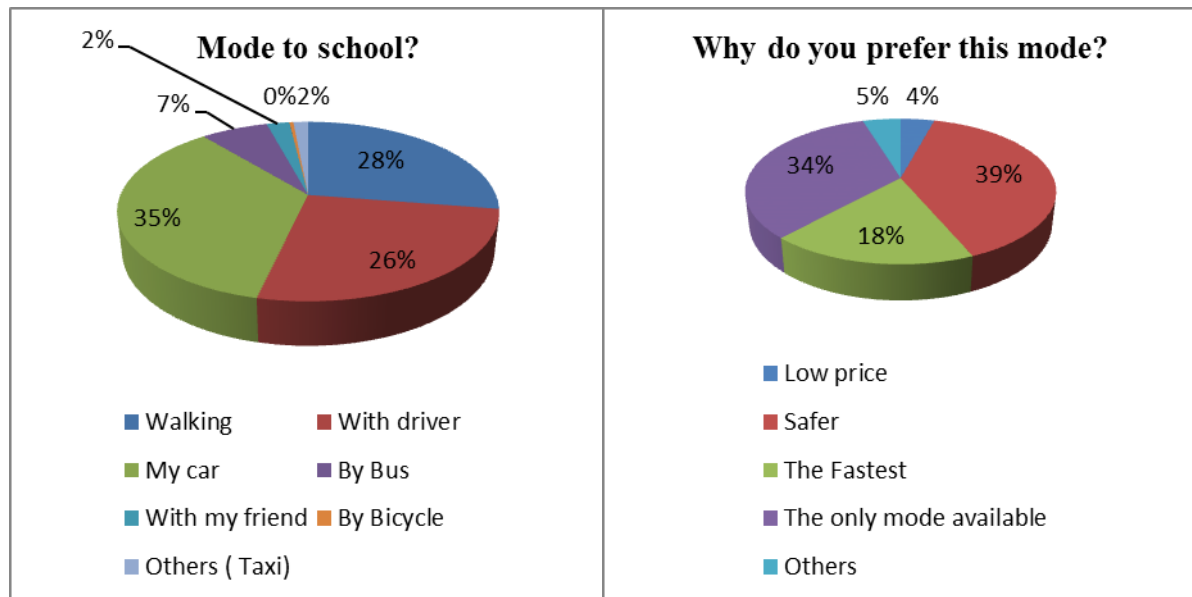


Figure 4.15: Mode to school

Figure 4.16: The reason for choosing this mode

From Figure 4.15, it is clear that 61% of the students of intermediary schools are coming to the schools by passenger cars.

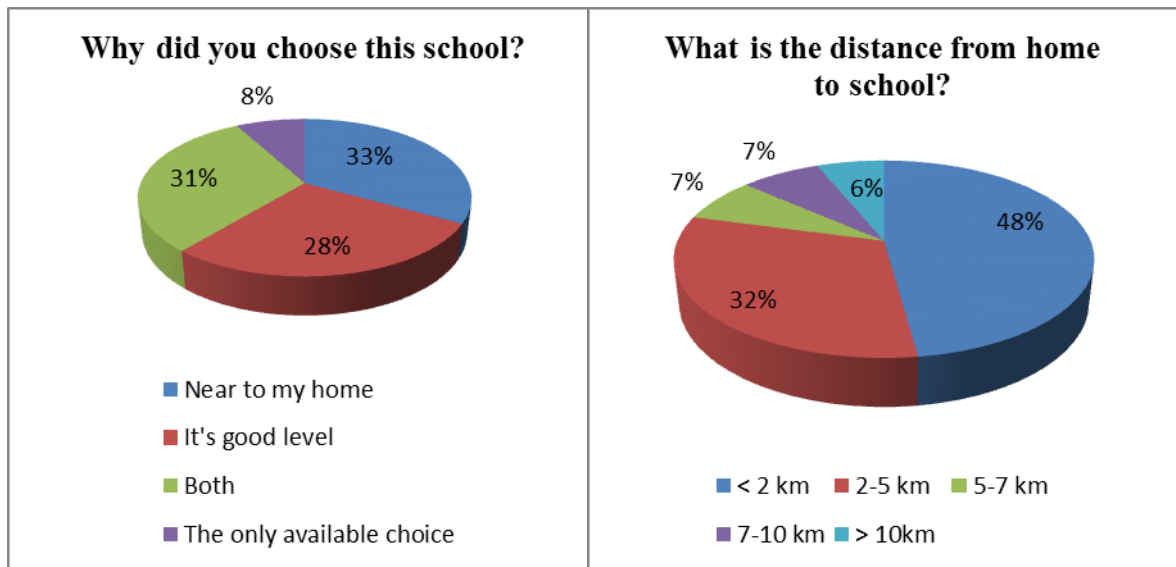


Figure 4.17: The reason for choosing this school

Figure 4.18: Distance from home to school

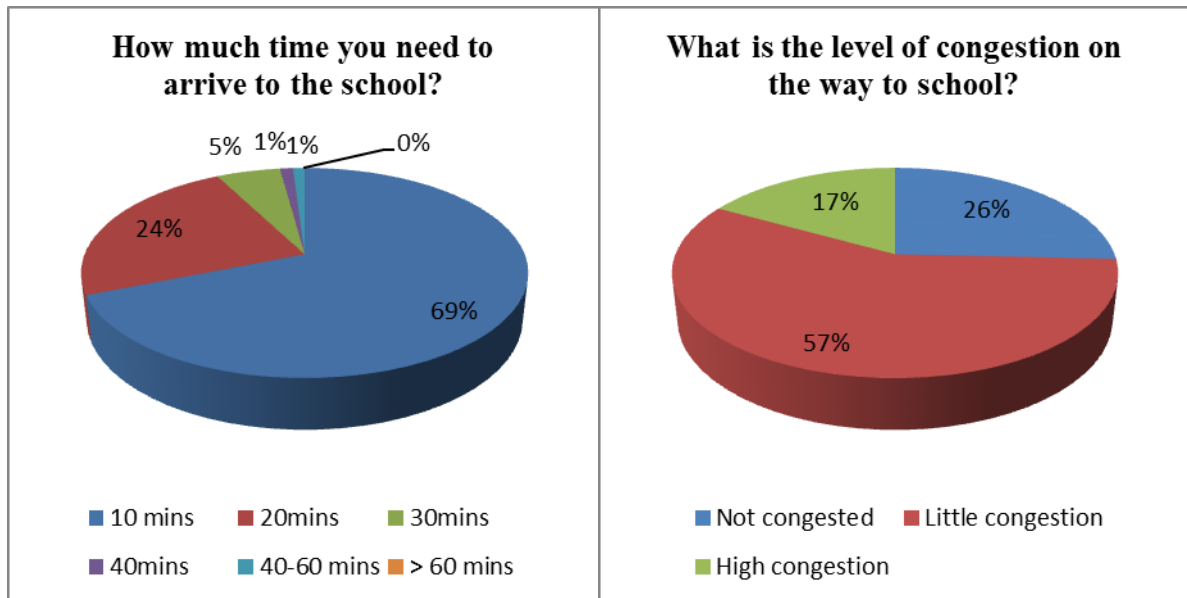


Figure 4.19: Time required to arrive to the school

Figure 4.20: Level of congestion on the way to school

From Figures 4.18 and 4.19, the average distance from home to school is 3.70 km, and the average time required to arrive to the school is 14 min.

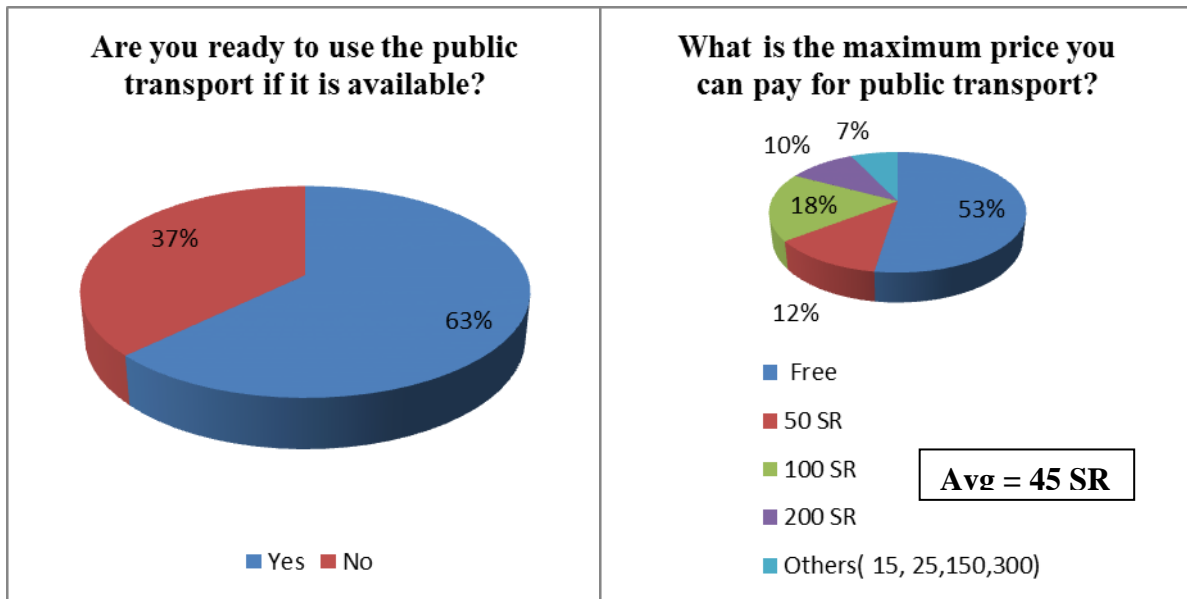


Figure 4.21: Willingness to use public transport

Figure 4.22: The maximum price for public transport

From Figures 4.21 and 4.22, 63% of the students of intermediary schools are ready to use public transportation, and 53% of them prefer this service to be free.

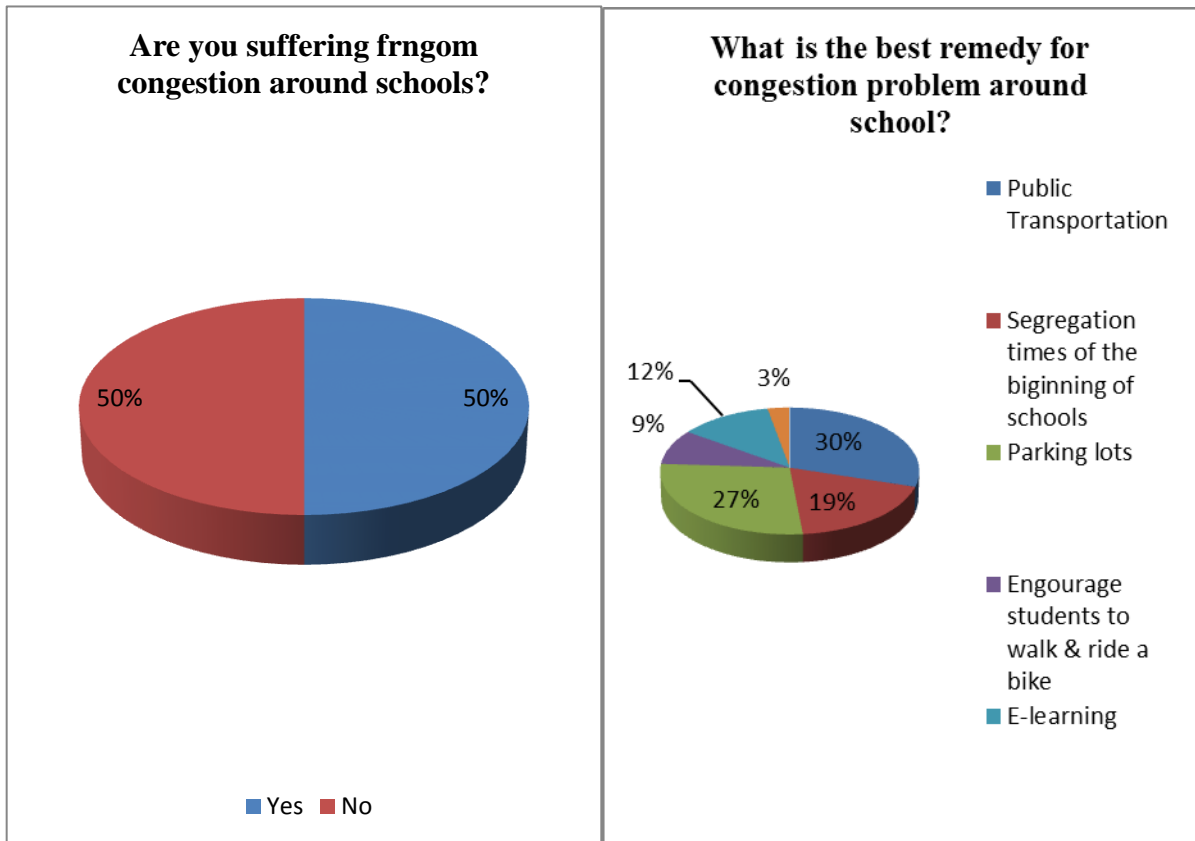


Figure 4.23: Suffering from congestion

Figure 4.24: Best remedies to solve congestion problem

According to the parents of students of intermediary schools, the best remedies to solve the congestion problem are providing the schools with more parking lots, segregation times of the beginning of schools, and public transportation, as shown in Figure 4.24.

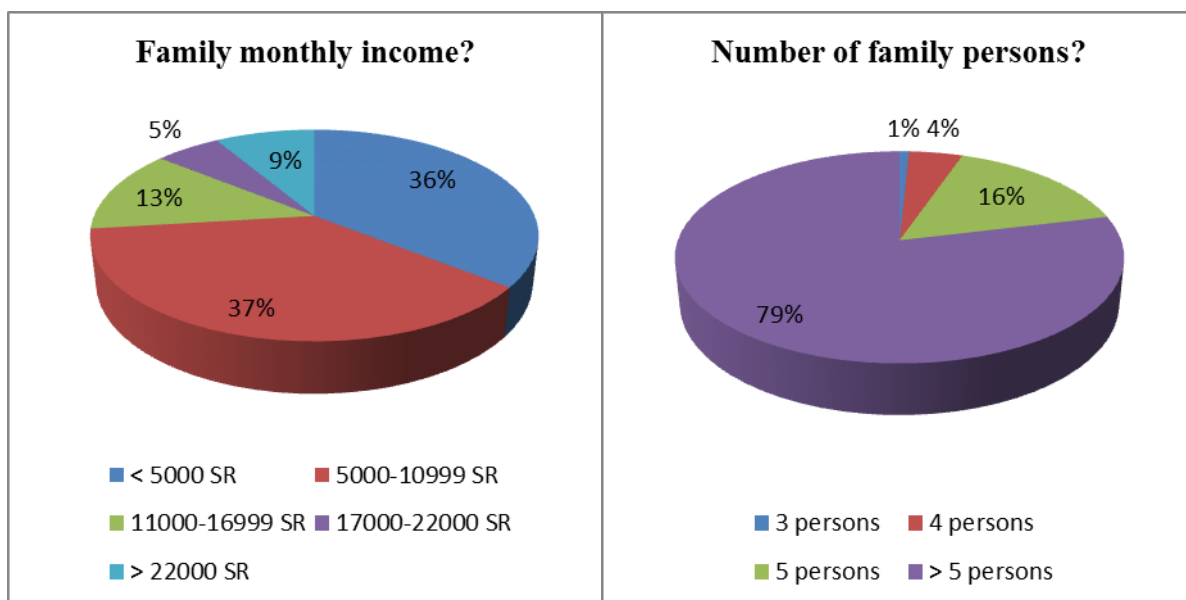


Figure 4.25: Family monthly income

Figure 4.26: Number of family persons

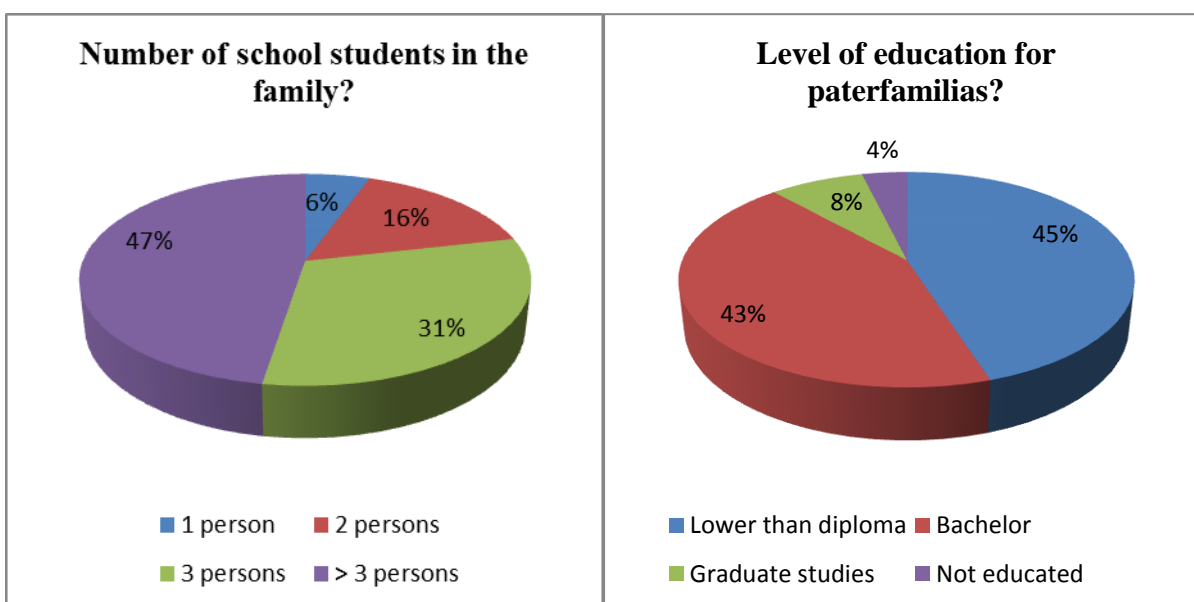


Figure 4.27: Number of students in the family

Figure 4.28: Level of education for paterfamilias

The weighted average for the family monthly income for students of intermediary schools is 9600 SR as shown in Figure 4.25.

4.3.3 Primary Schools

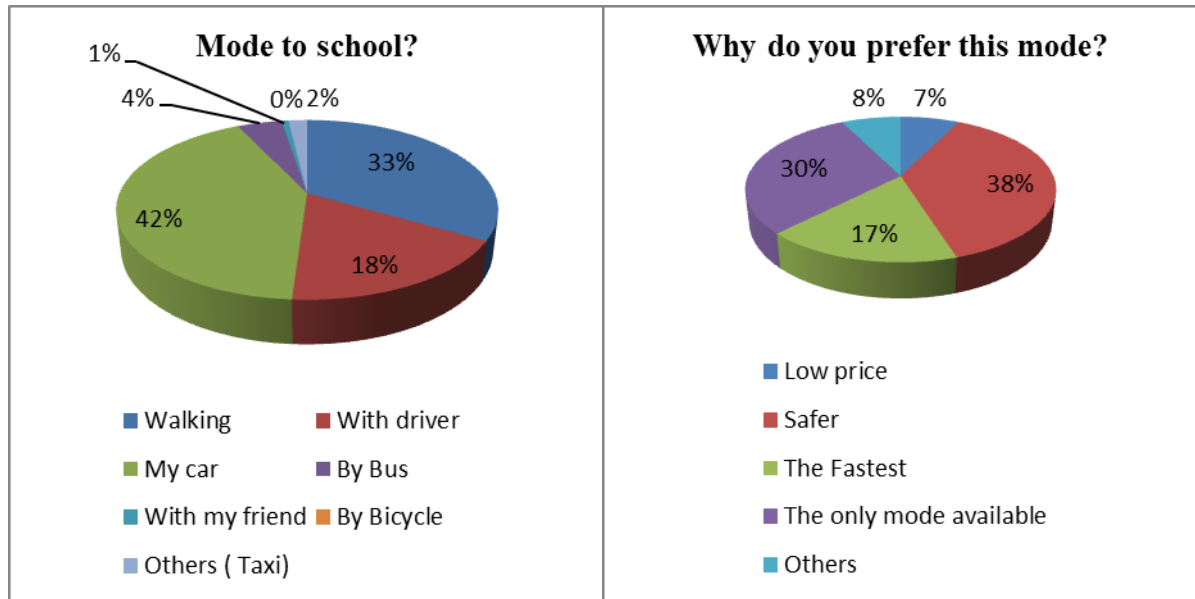


Figure 4.29: Mode to school

Figure 4.30: The reason for choosing this mode

From Figure 4.29, it is clear that 60% of the students of primary schools are coming to the schools by passenger cars.

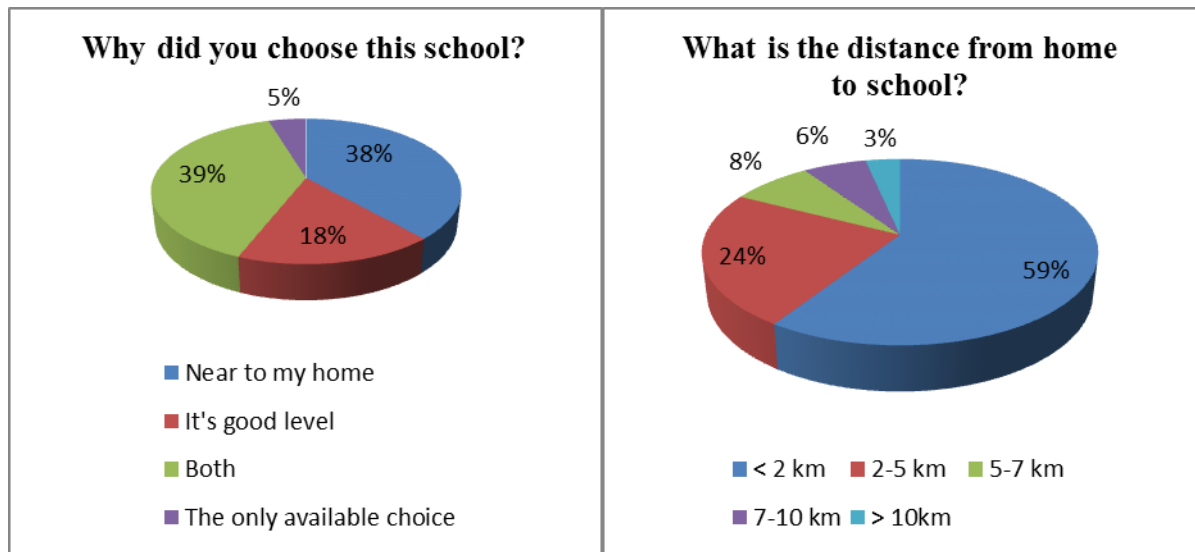


Figure 4.31: The reason for choosing this school

Figure 4.32: Distance from home to school

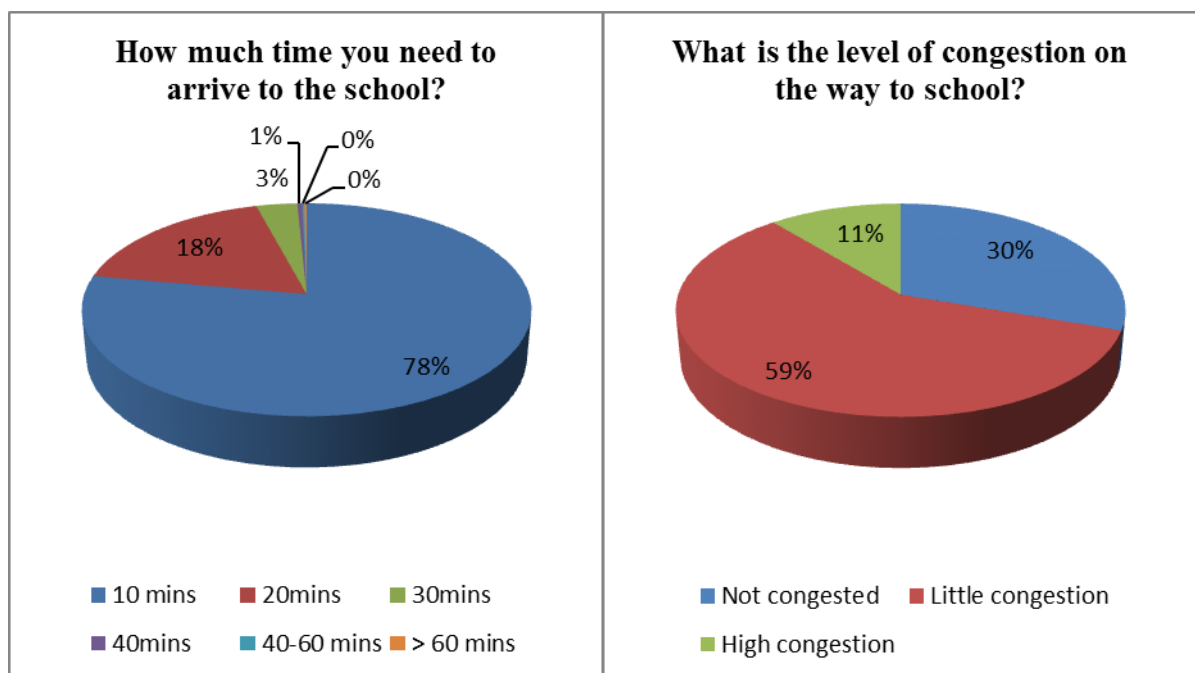


Figure 4.33: Time required to arrive to the school

Figure 4.34: Level of congestion on the way to school

From Figures 4.32 and 4.33, the average distance from home to school is 3.30 km, and the average time required to arrive to the school is 12 min.

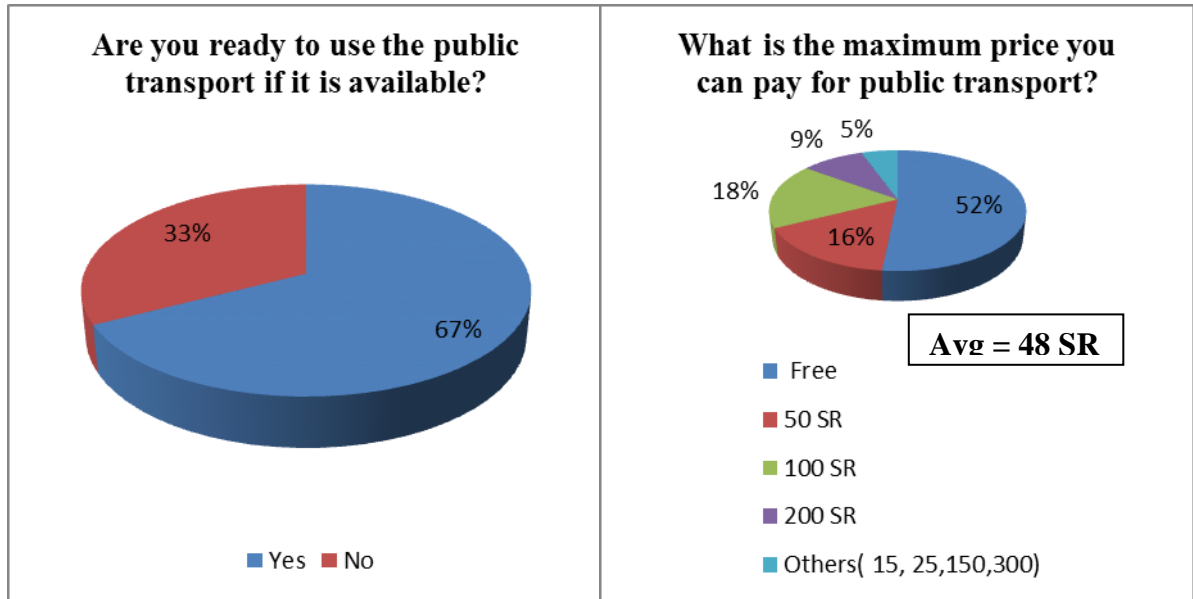


Figure 4.35: Willingness to use public transport

Figure 4.36: The maximum price for public transport

From Figures 4.35 and 4.36, 67% of the students of primary schools are ready to use public transportation, and 52% of them prefer this service to be free.

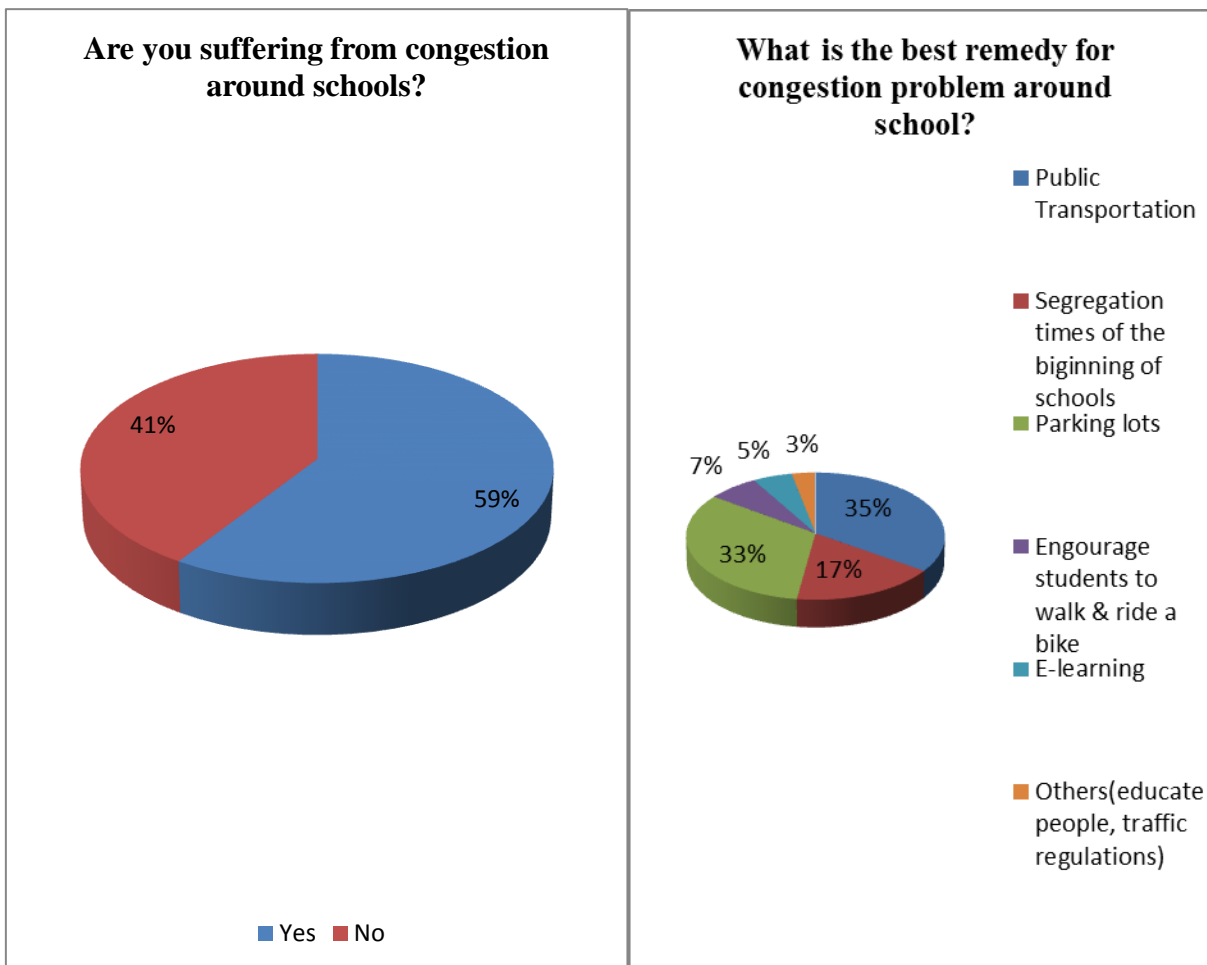


Figure 4.37: Suffering from congestion

Figure 4.38: Best remedies to solve congestion problem

According to the parents of students of primary schools, the best remedies to solve the congestion problem are providing the schools with more parking lots, segregation times of the beginning of schools, and public transportation, as shown in Figure 4.38.

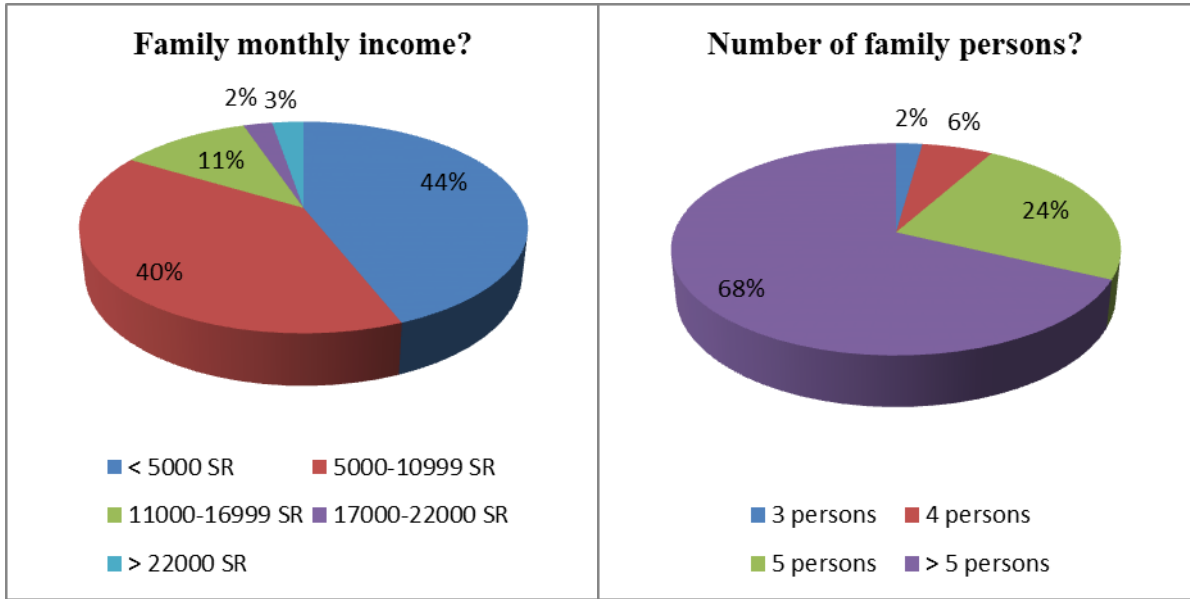


Figure 4.39: Family monthly income

Figure 4.40: Number of family persons

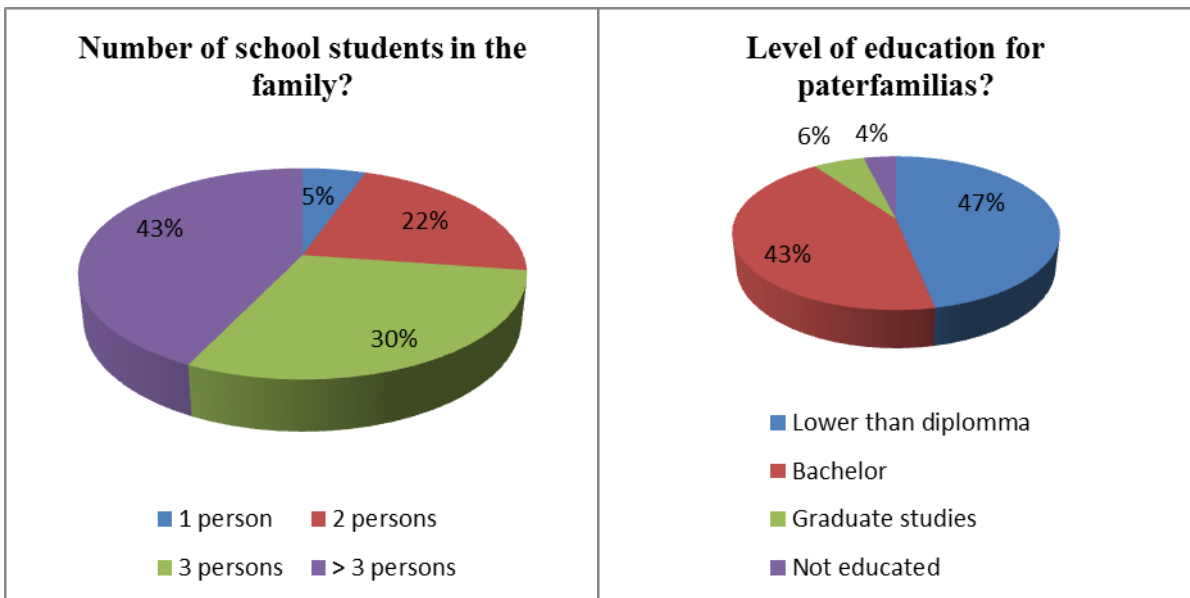


Figure 4.41: Number of students in the family

Figure 4.42: Level of education for paterfamilias

The weighted average for the family monthly income for students of primary schools is 8000 SR as shown in Figure 4.39.

4.3.4 All Levels

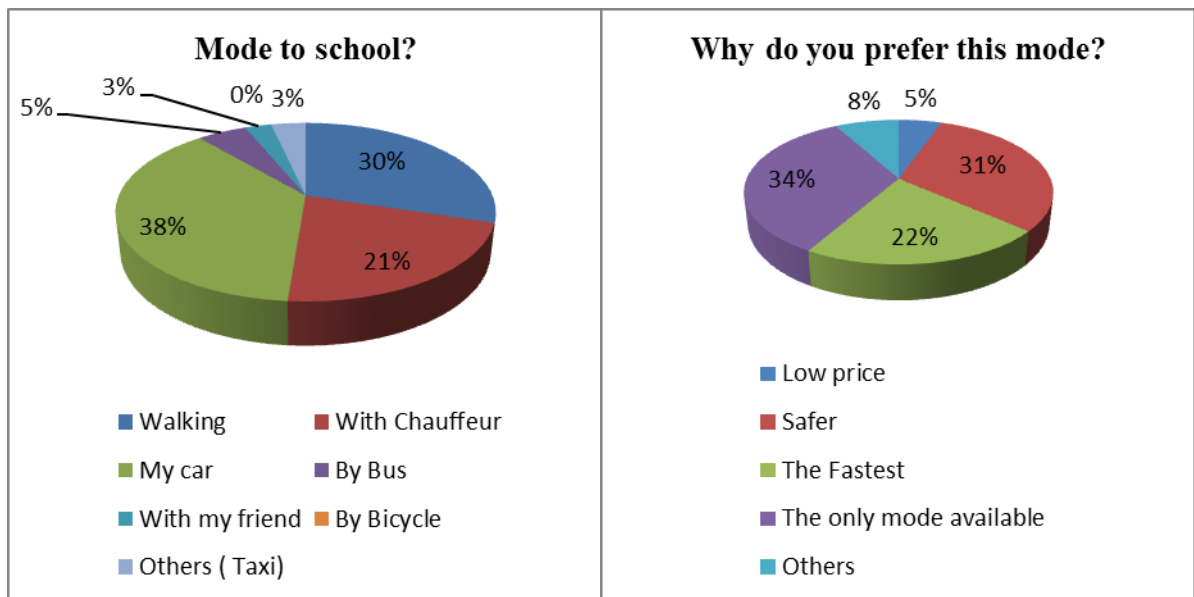


Figure 4.43: Mode to school

Figure 4.44: The reason for choosing this mode

From Figure 4.43, it is clear that 59% of the students of all schools are coming to the schools by passenger cars.

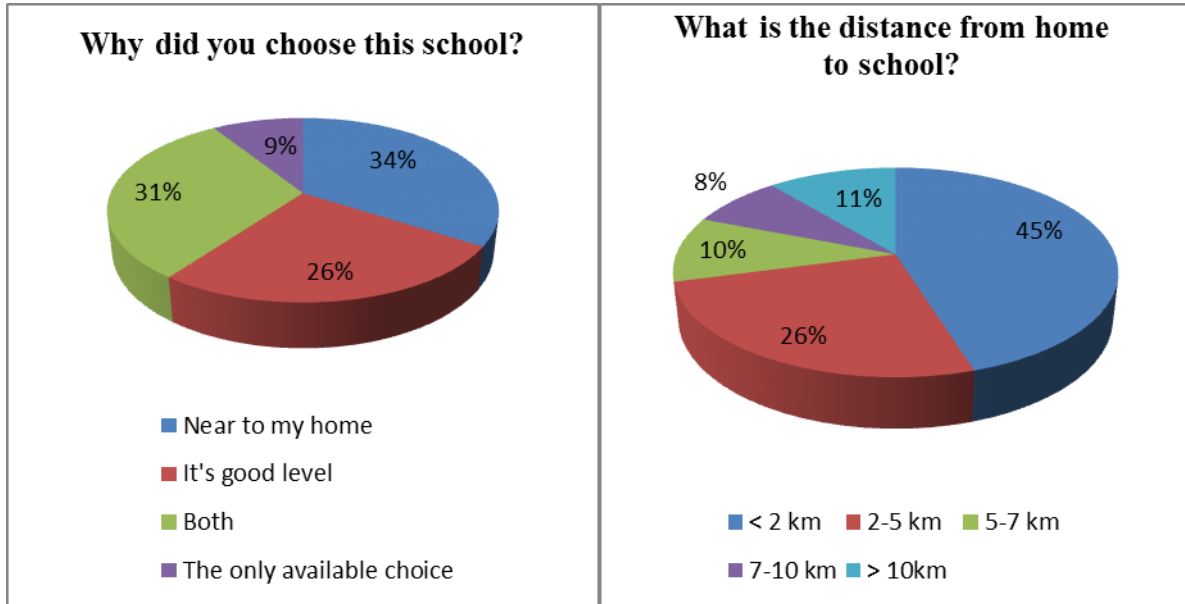


Figure 4.45: The reason for choosing this school

Figure 4.46: Distance from home to school

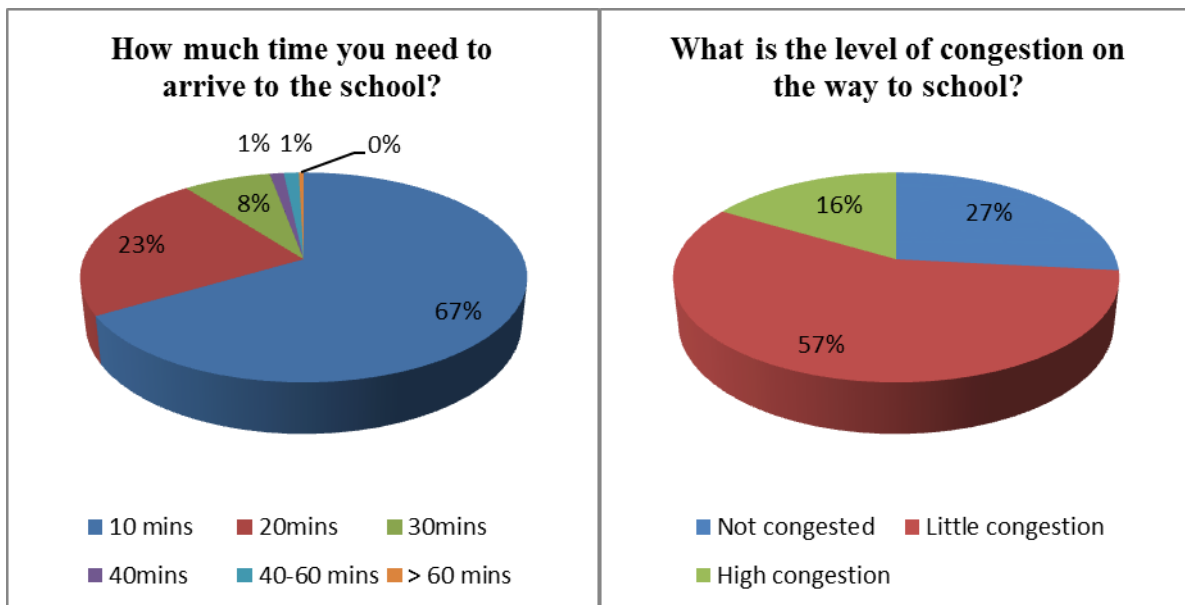


Figure 4.47: Time required to arrive to the school

Figure 4.48: Level of congestion on the way to school

From Figures 4.46 and 4.47, the average distance from home to school is 4.00 km, and the average time required to arrive to the school is 14 min.

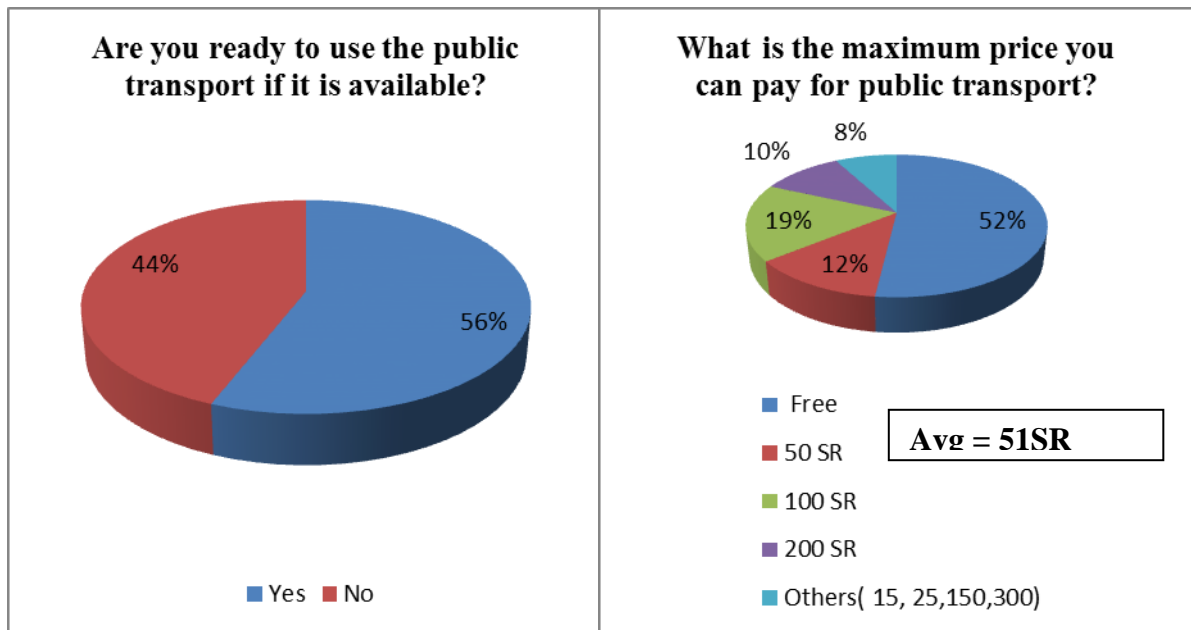


Figure 4.49: Willingness to use public transport

Figure 4.50: The maximum price for public transport

From Figures 4.49 and 4.50, 67% of all students are ready to use public transportation, and 52% of them prefer this service to be free.

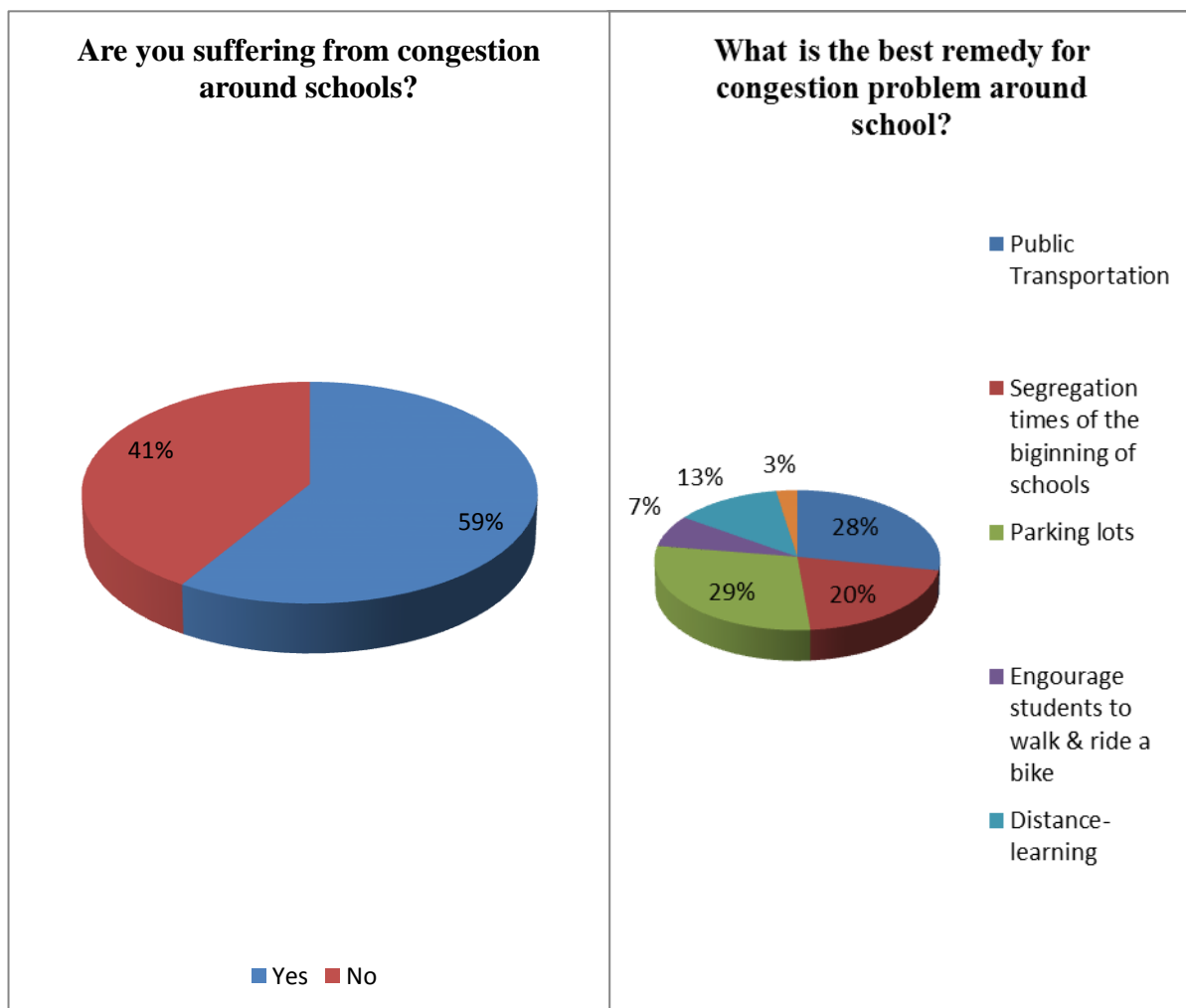


Figure 4.51: Suffering from congestion

Figure 4.52: Best remedies to solve congestion problem

According to the parents and students, the best remedies to solve the congestion problem are providing the schools with more parking lots, segregation times of the beginning of schools, and public transportation, as shown in Figure 4.52.

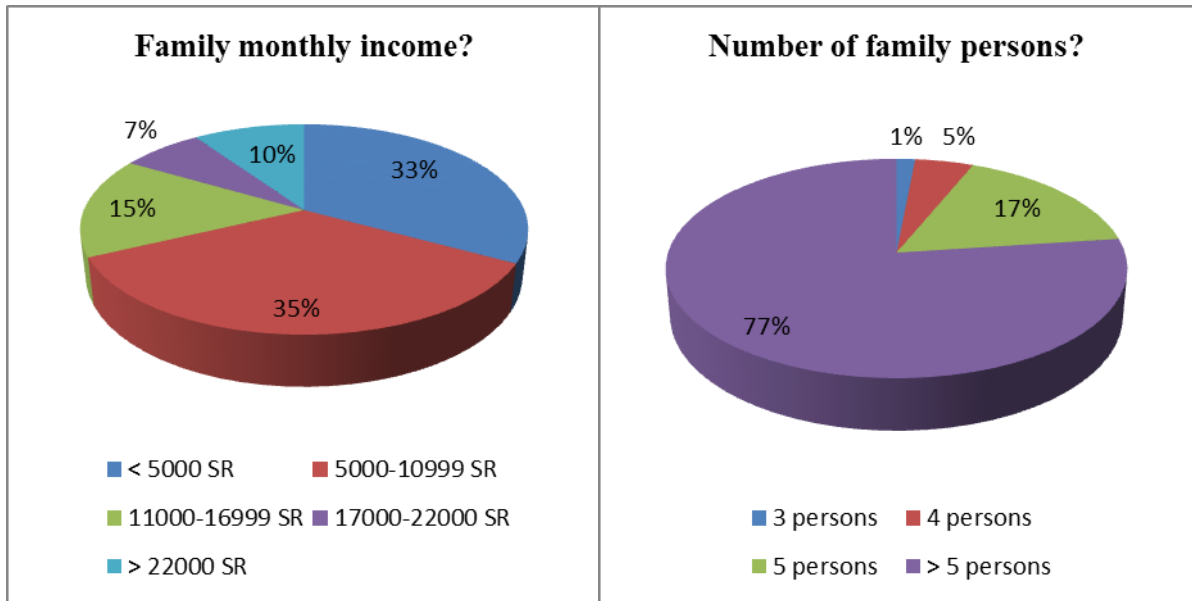


Figure 4.53: Family monthly income

Figure 4.54: Number of family persons

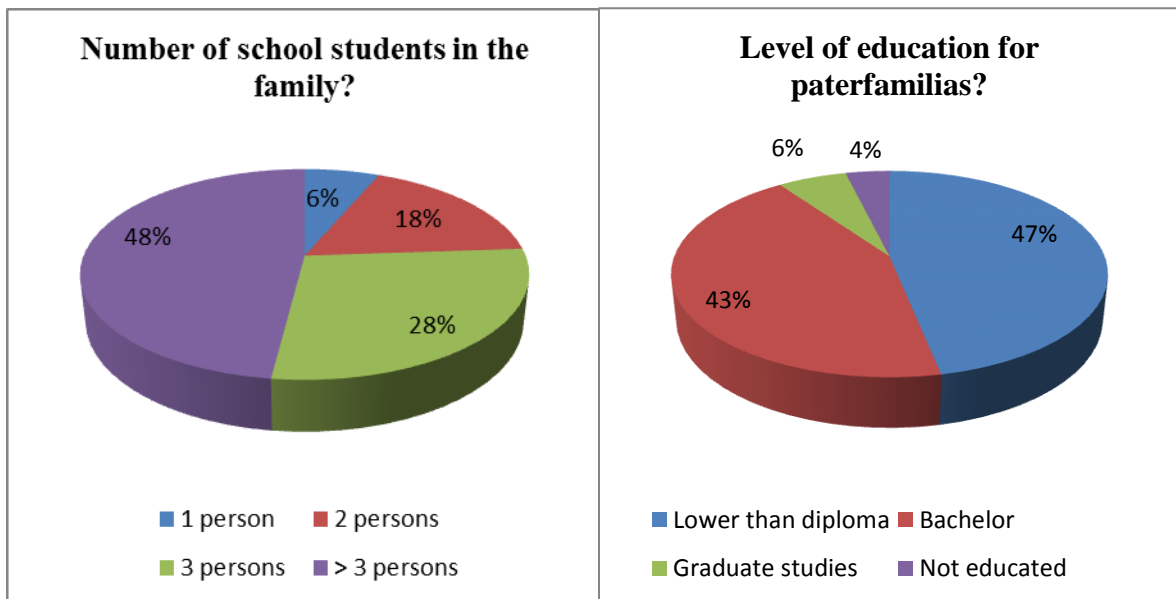


Figure 4.55: Number of students in the family

Figure 4.56: Level of education for paterfamilias

The weighted average for the family monthly income for all students is 10000 SR as shown in Figure 4.53.

4.4 Summary of Results

From the previous discussion and charts, it is very obvious that although nearly 60% of the students are suffering from congestion around schools, still more than 55% of the students are still coming to schools by passenger cars. On average, 56% of all students are ready to use public transportation if it is free service. The average time that is required from all the students of all levels to arrive to the school is 14 min. Nearly 28% of the respondents suggested that enhancing public transportation is the best remedy to overcome the congestion problem around schools. Another suggestion which was presented by 29% of the respondents is providing the schools with enough area for parking lots. The remaining respondents suggested segregation of school times at the beginning of the schools, encourage students to walk, strict enforcement of traffic regulations, and application of distance learning. Correlation matrices were developed to show the relationship between the number of attracted and parked vehicles and the continuous variables of the questionnaire, such as distance to school, time to school, monthly family income, number of family members, and the number of students in the family. The matrices are shown in Tables 4.5 to 4.8.

Table 4.5: Correlation with the new variables for secondary schools

	<i>Parked</i>	<i>Total Attracted</i>	<i>Distance to home</i>	<i>Time to school</i>	<i>Family income</i>	<i>Family members</i>	<i>Students in family</i>
Parked	1.00						
Total Attracted	0.80	1.00					
Distance to home	-0.44	-0.52	1.00				
Time to school	-0.43	-0.35	0.74	1.00			
Family income	0.45	-0.13	0.24	-0.20	1.00		
Family members	0.92	0.51	-0.23	-0.40	0.76	1.00	
Students in family	0.76	0.61	-0.57	-0.60	0.35	0.71	1.00

Table 4.6: Correlation with the new variables for intermediary schools

	<i>Parked</i>	<i>Total Attracted</i>	<i>Distance to school</i>	<i>Time to school</i>	<i>Family income</i>	<i>Family members</i>	<i>Students in family</i>
Parked	1.00						
Total Attracted	0.72	1.00					
Distance to school	0.57	0.61	1.00				
Time to school	0.36	0.01	0.52	1.00			
Family income	0.18	0.26	0.22	-0.03	1.00		
Family members	-0.67	-0.32	0.18	-0.17	0.10	1.00	
Students in family	-0.67	-0.32	0.18	-0.17	0.10	1.00	1.00

Table 4.7: Correlation with the new variables for primary schools

	<i>Parked</i>	<i>Total Attracted</i>	<i>Distance to school</i>	<i>Time to school</i>	<i>Family income</i>	<i>Family members</i>	<i>Students in family</i>
Parked	1.00						
Total Attracted	0.92	1.00					
Distance to home	0.14	0.48	1.00				
Time to school	-0.04	0.30	0.80	1.00			
Family income	0.26	0.51	0.86	0.39	1.00		
Family members	0.88	0.77	-0.14	-0.05	-0.14	1.00	
Students in family	0.69	0.77	0.21	0.19	0.23	0.79	1.00

Table 4.8: Correlation with the new variables for all levels schools

	<i>Parked</i>	<i>Total Attracted</i>	<i>Distance to home</i>	<i>Time to school</i>	<i>Family income</i>	<i>Family members</i>	<i>Students in family</i>
Parked	1.00						
Total Attracted	0.39	1.00					
Distance to home	0.44	0.02	1.00				
Time to school	0.50	0.13	0.82	1.00			
Family income	0.50	0.25	0.53	0.44	1.00		
Family members	0.59	0.23	0.21	0.18	0.33	1.00	
Students in family	0.07	0.26	-0.27	-0.19	0.10	0.66	1.00

From the correlation matrices above, there are strong relationships between the number of attracted and parked vehicles and the number of family members and number of students in the family when the school level is secondary or primary since the correlation coefficients are very high. For intermediary schools, there is a strong relationship between the number of attracted and parked vehicles and the distance to school. It is recommended to use these new variables which are significant in modeling of trip attraction and parking generation.

CHAPTER 5

CASE STUDY

5.1 Introduction

Optimization of the overall network during the peak hour of the school has good effect in reducing congestion around schools. This chapter concentrates on the effect of optimizing an intersection near a primary school and demonstrates that a straightforward engineering solution can have positive impacts on congestion around schools. This study was done at the intersection of Prince Miqrin Street and 22nd Street near Al-Seddeeq primary school in Al-Aqrabiyya. The gate of the school is located on the west leg of the intersection as shown in Figure 5.1 below.

From the field visits of the school, it was obvious that during the AM peak hour of the school, a lot of delay happened for most of the vehicles using the intersection since 45 sec was allocated as green time for eastbound vehicles (west leg) and only 60 sec of the green time was for the other bounds. A team of 4 persons counted the directional number of vehicles using the intersection from 6:00 AM to 7:30 AM.

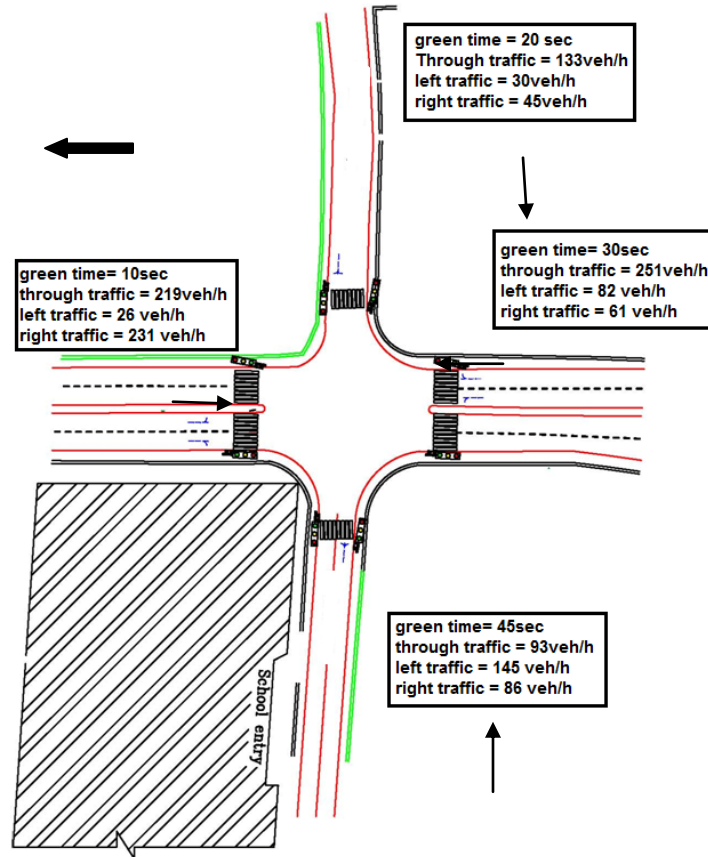


Figure 5.1: Intersection layout

5.2 Data Collection

The number of vehicles with their directions was collected using the mechanical counters from 6:00 AM to 7:30 AM. The cycle length for the intersection was 125 sec. 45 sec of green was for eastbound, 10 sec for southbound, 20 sec for westbound, and 30 sec for the northbound. The number of vehicles for each 15 minutes is presented in Tables 5.1 to 5.4.

Table 5.1: Traffic counts for eastbound

Starting time	Eastbound			Total vehicles
	P			
	R	S	L	
6:00-6:15 AM	7	4	9	20
6:15-6:30 AM	13	14	20	47
6:30-6:45 AM	17	24	49	90
6:45-7:00 AM	23	34	43	100
7:00-7:15 AM	33	21	33	87
7:15-7:30 AM	20	26	17	63
Total	113	123	171	407

Table 5.2: Traffic counts for southbound

Starting time	Southbound			Total vehicles
	P			
	R	S	L	
6:00-6:15 AM	15	6	2	23
6:15-6:30 AM	36	29	5	70
6:30-6:45 AM	79	64	5	148
6:45-7:00 AM	61	77	8	146
7:00-7:15 AM	55	49	8	112
7:15-7:30 AM	23	25	5	53
Total	269	250	33	552

Table 5.3: Traffic counts for westbound

Starting time	Westbound			Total vehicles
	P			
	R	S	L	
6:00-6:15 AM	5	11	7	23
6:15-6:30 AM	10	14	5	29
6:30-6:45 AM	10	35	5	50
6:45-7:00 AM	16	38	12	66
7:00-7:15 AM	9	46	8	63
7:15-7:30 AM	7	26	13	46
Total	57	170	50	277

Table 5.4: Traffic counts for northbound

Starting time	Northbound			Total vehicles
	P			
	R	S	L	
6:00-6:15 AM	7	19	5	31
6:15-6:30 AM	15	51	14	80
6:30-6:45 AM	15	73	24	112
6:45-7:00 AM	15	75	25	115
7:00-7:15 AM	16	52	19	87
7:15-7:30 AM	8	37	17	62
Total	76	307	104	487

The peak hour was from 6:15-7:15 AM since the maximum volume came at this hour.

5.3 Simulation and Optimization

TRANSYT-7F was used to simulate the existing conditions and then used to optimize the best conditions in which the delay time and number of stops were minimal. Table 5.5 shows the performance measures for the existing conditions while Table 5.6 shows the same measures for the optimal conditions. The objective function was to minimize the fuel consumption.

Table 5.5: Simulation of the existing conditions

Cycle Length = 125 sec		
Green time for southbound = 10sec		
Green time for westbound = 20 sec		
Green time for northbound = 30 sec		
Green time for eastbound = 45 sec		
Performance Measures	Units	System Totals
Total Travel Time	veh-km/hr	144
Total Uniform Delay	veh-hr/hr	47
Total Random Delay	veh-hr/hr	87
Total Delay	veh-hr/hr	134
Average Delay	sec/veh	282.8
Passenger Delay	pax-hr/hr	161
Uniform Stops	veh/hr	1596
Random Stops	veh/hr	1228
Total Stops	veh/hr	2824
Degree of Sat > 1	# of links	2
Queue Spillback	# of links	0
Time Jammed	%	0
System Speed	km/hr	4.8
Fuel Consumption	lit/hr	502
Operating Cost	\$/hr	716
Performance Index	DI	133.10

Table 5.6: Simulation of the optimal conditions

Cycle Length = 100 sec		
Green time for southbound = 24sec		
Green time for westbound = 15 sec		
Green time for northbound = 13 sec		
Green time for eastbound = 28 sec		
Performance Measures	Units	System Totals
Total Travel Time	veh-km/hr	67
Total Uniform Delay	veh-hr/hr	22
Total Random Delay	veh-hr/hr	36
Total Delay	veh-hr/hr	58
Average Delay	sec/veh	121.9
Passenger Delay	pax-hr/hr	69
Uniform Stops	Veh/hr	1930
Random Stops	Veh/hr	956
Total Stops	Veh/hr	2886
Degree of Sat > 1	# of links	3
Queue Spillback	# of links	0
Time Jammed	%	0
System Speed	km/hr	10.2
Fuel Consumption	lit/hr	306
Operating Cost	\$/hr	544
Performance Index	DI	77.90

5.4 Summary

By comparing the performance measures for the existing conditions and optimal conditions, it is clear that the average delay for each vehicle was reduced by more than 50%, which led to the reduction in fuel consumption and operating cost and increased the system speed from 4.8 km/hr to 10.2 km/hr. Furthermore, it is clear that the performance index represented by the excess fuel consumption was reduced dramatically from 133.1 to 77.90, which means that some simple engineering remedies can make a lot of positive impacts. The existing green time for the eastbound leg is 45 sec but the optimized one is 28 sec. The designer of the signal timing thought that the existence of the school on the eastbound leg will increase the traffic volume on that leg, but in actual conditions most of the traffic in the AM peak hour was for north and south bounds in which the green time for both of them was 40 sec, resulting in increase in the delay and fuel consumption.

CHAPTER 6

CONCLUSIONS AND RECOMMENDATIONS

The ultimate objective of this study was to evaluate the traffic impact of schools and this was accomplished in three parts which are: modeling of trip and parking generation, assessing the acceptance by students and their parents of some of the remedies for mitigating congestion around schools through questionnaire survey, and an engineering exercise to show and prove that a simple straightforward engineering solution can have a considerable positive impact. From the analyses and results for these three parts, it can be concluded that:

- It is obvious that the worst models in predicting the number of attracted vehicles to schools are Riyadh city manual models since in most cases these models overestimated the number of attracted vehicles, especially when the variable used is the lot or floor area.
- ITE and Dubai trip generation rates predicted the number of attracted vehicles much better than that predicted by Riyadh city rates.
- The best variables in predicting the number of attracted and parked vehicles were the number of students and number of employees. Tables 6.1 to 6.2 present the best models for trip and parking generation for each level of schools.

Table 6.1: Best models for trip attraction of each level of schools

School level	Best models (trip attraction)	R ²	*Planning model
Secondary	Attracted vehicles = 4.84 * Number of employees	0.86	Attracted vehicles = 36.47 * Floor area
Intermediary	Attracted vehicles = 0.56 * Number of students	0.87	Attracted vehicles = 57 * Floor area
Primary	Attracted vehicles = 5.40 * Number of employees	0.85	Attracted vehicles = 61 * Floor area

**Can be used in planning stage when detail data about schools is not available.*

Table 6.2: Best models for parking generation of each level of schools

School level	Best models (parking generation)	R ²	* Planning model
Secondary	Parked vehicles = 1.97 * Number of employees	0.82	Parked vehicles = 14.76 * Floor area
Intermediary	Parked vehicles = 0.08 * Number of students	0.87	Parked vehicles = 8.5 * Floor area
Primary	Parked vehicles = 0.95 * Number of employees	0.89	Parked vehicles = 10.5 * Floor area

**Can be used in planning stage when detail data about schools is not available.*

- In planning stage in which the information is scarce, using the floor area is adequate but with less accuracy.
- The worst variable in predicting the number of attracted and parked vehicles was the lot area.
- It is better to deal with each level of schools separately since the models that resulted from the analysis of pooled schools together are statistically less significant from the others.

- It is suggested to use the developed models rather than the local models or international models in the study area.
- In other areas of the Kingdom, the planner has an option to use the developed models or the ITE models since there is no big difference between the predicted values using both models.
- From the questionnaire analysis, it is obvious that 60% of overall students are suffering from congestion around schools but still 59% of them are coming to schools by passenger cars.
- 33% of overall students are ready to use the public transportation if it is free service and 27% are ready to pay an average of 51 SR.
- 30% of the people suggested public transportation as a solution for congestion, 28% suggested providing more parking lots, 18% suggested segregation time as a remedy, 12% suggested distance learning, and others suggested encouraging students to walk and educate people.
- As shown in the case study, some simple engineering improvement such as optimization of signal timing can make a big difference in reducing congestion.

CHAPTER 7

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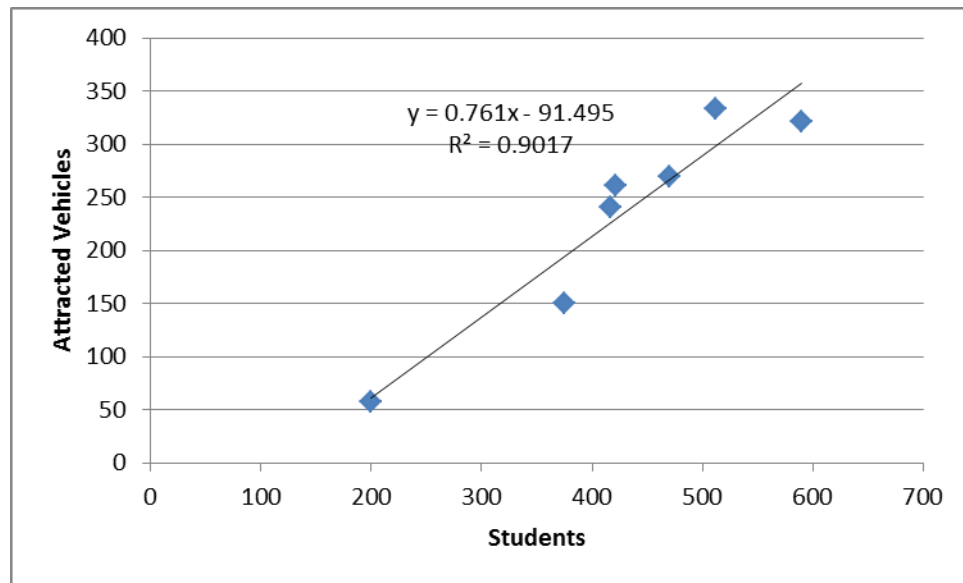
APPENDIX A

SAMPLE OF DATA ANALYSIS FOR TRIP ATTRACTION (INTERMEDIARY SCHOOLS)

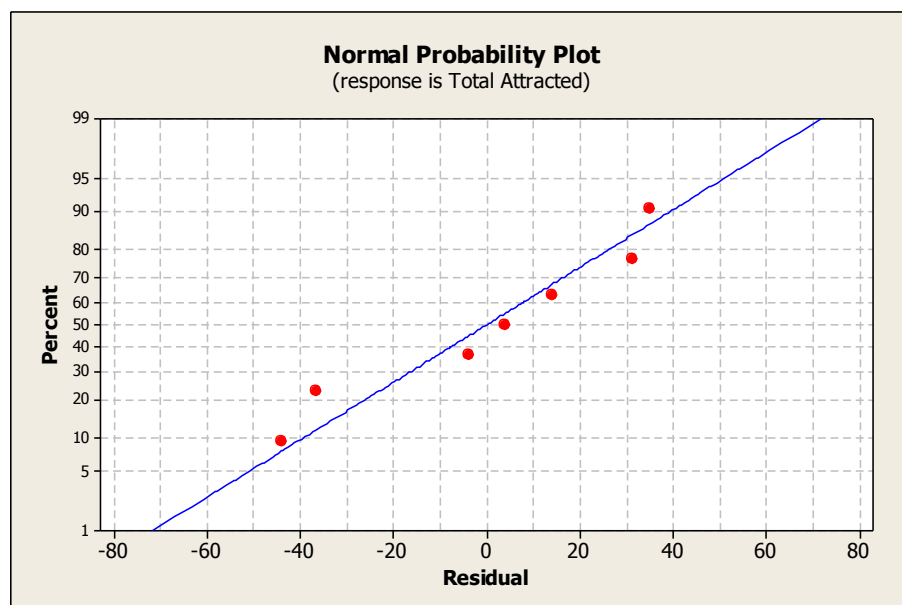
Data analysis using seven schools only

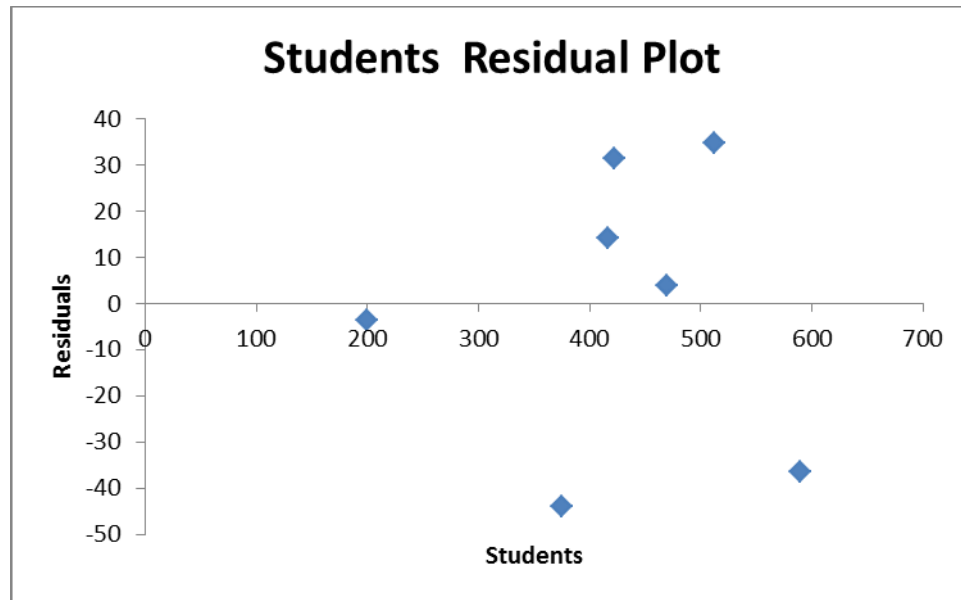
Simple linear regression

Students



Correlation of the number of students with the volume attracted (school level: Secondary)





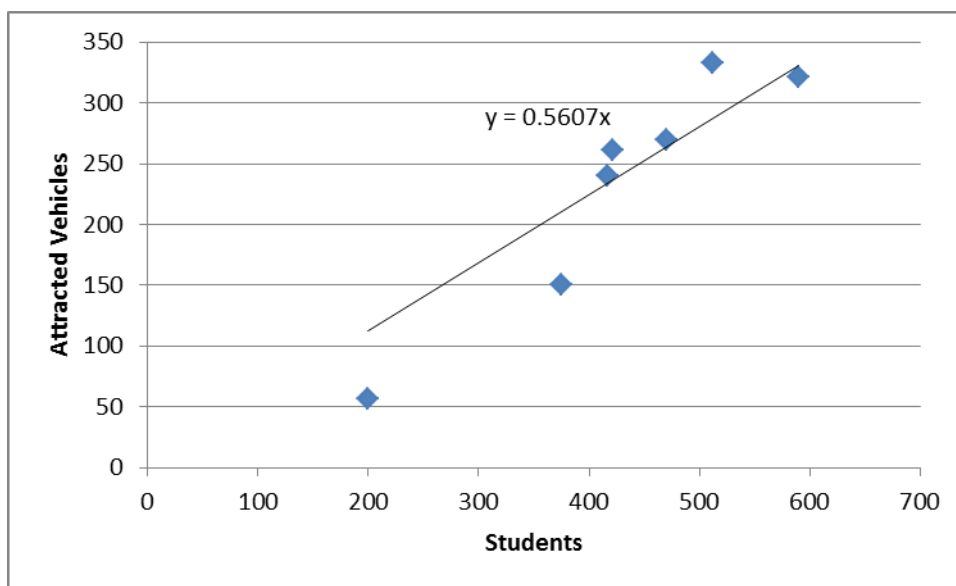
SUMMARY OUTPUT

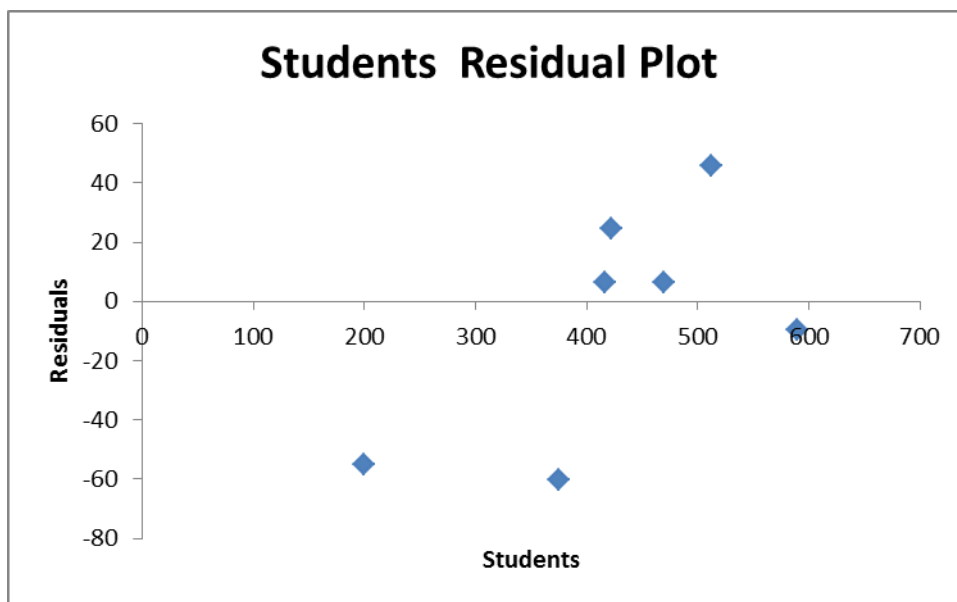
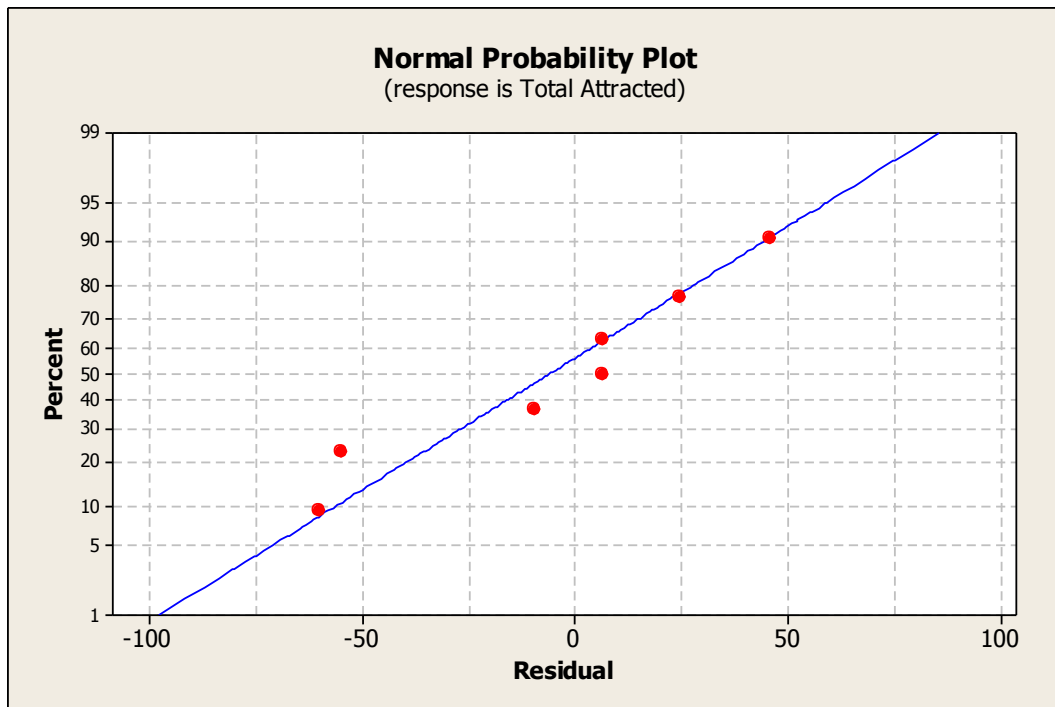
<i>Regression Statistics</i>	
Multiple R	0.949563
R Square	0.901669
Adjusted R Square	0.882003
Standard Error	33.71819
Observations	7

<i>ANOVA</i>					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	52126.27	52126.27	45.84881	0.001067818
Residual	5	5684.583	1136.917		
Total	6	57810.86			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
	-			
Intercept	91.49528094	49.60904147	-1.84433	0.124453
Students	0.761040511	0.112394161	6.771175	0.001068

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





SUMMARY OUTPUT

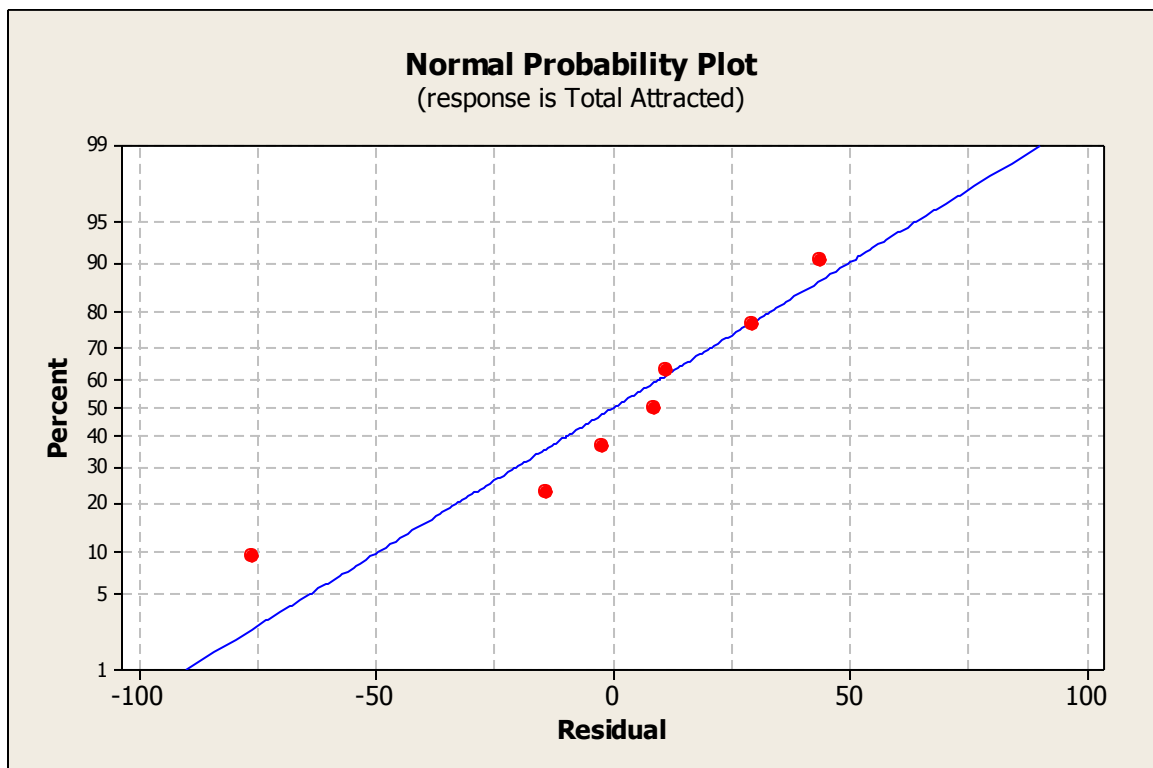
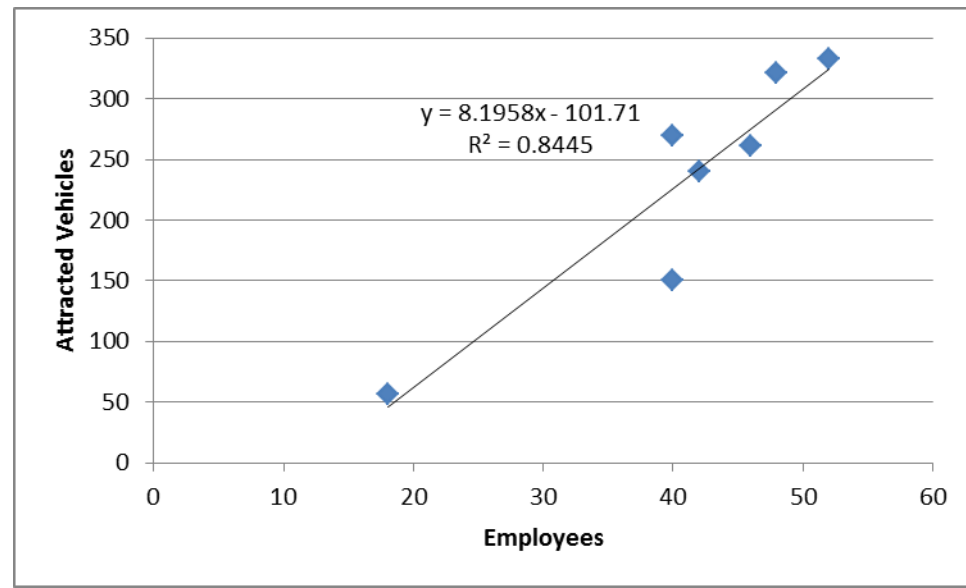
<i>Regression Statistics</i>	
Multiple R	0.989044
R Square	0.978207
Adjusted R Square	0.81154
Standard Error	39.89957
Observations	7

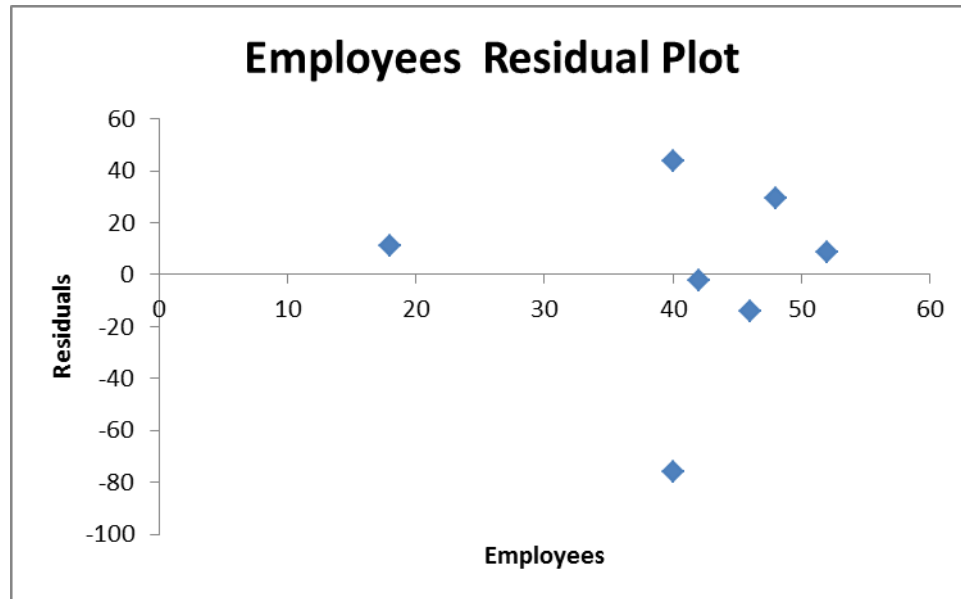
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	428748.1	428748.1	269.3183	1.5329E-05
Residual	6	9551.852	1591.975		
Total	7	438300			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Students	0.56070576	0.034166623	16.41092	3.26E-06

Employees





SUMMARY OUTPUT

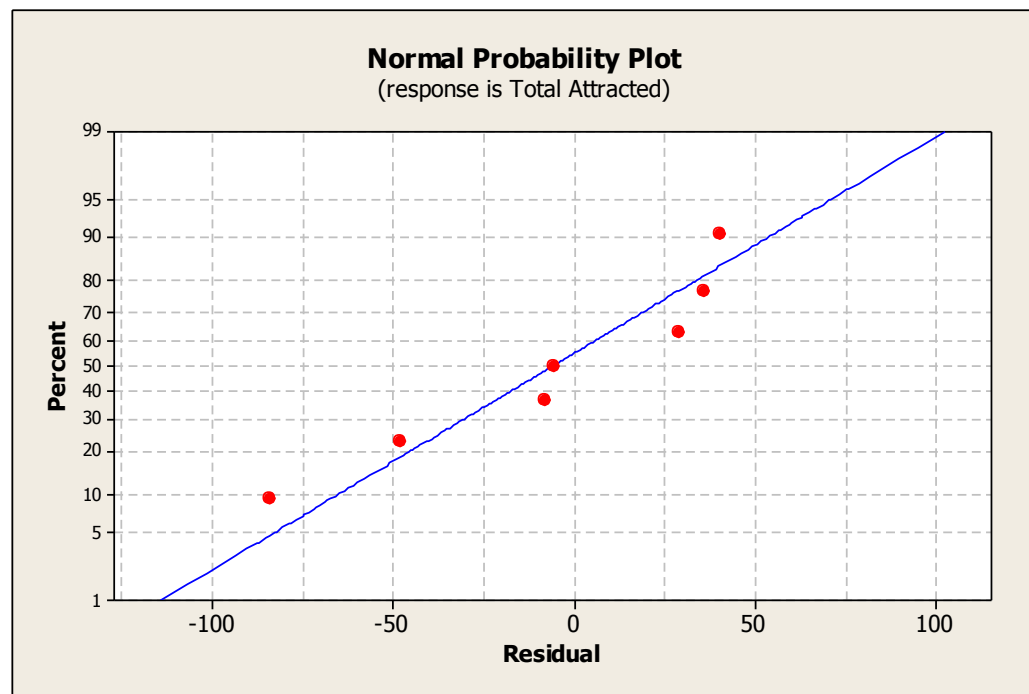
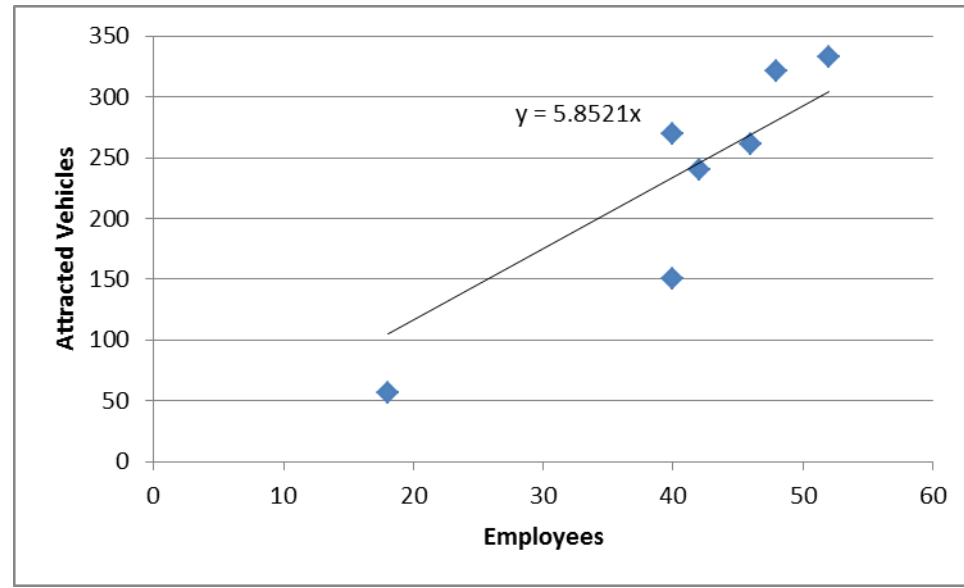
<i>Regression Statistics</i>	
Multiple R	0.918985755
R Square	0.844534817
Adjusted R Square	0.813441781
Standard Error	42.39711186
Observations	7

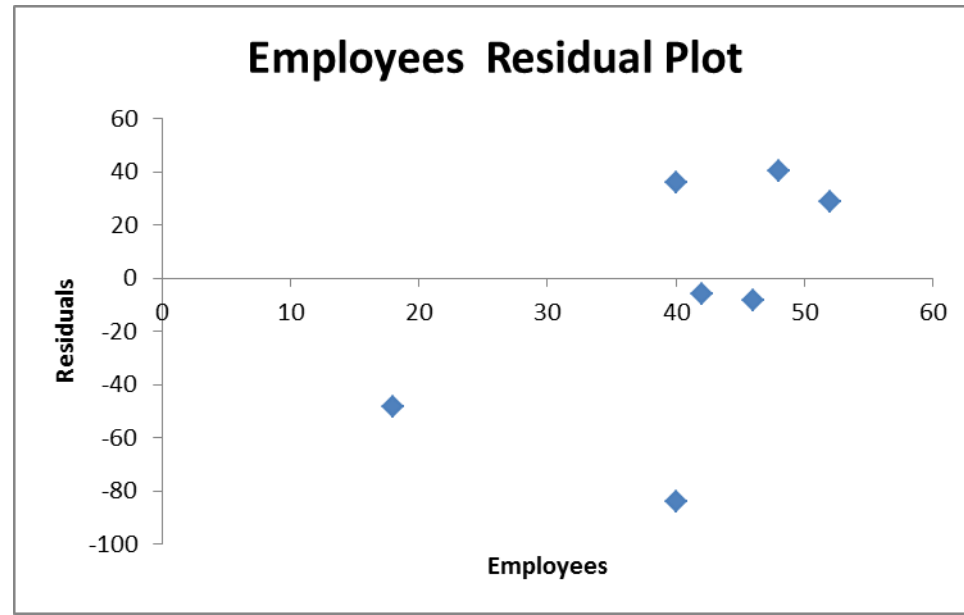
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	48823.28	48823.28	27.16154	0.003433551
Residual	5	8987.575	1797.515		
Total	6	57810.86			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	-101.7122642	66.21915051	-1.53599	0.185138
Employees	8.195754717	1.572576356	5.211674	0.003434

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





SUMMARY OUTPUT

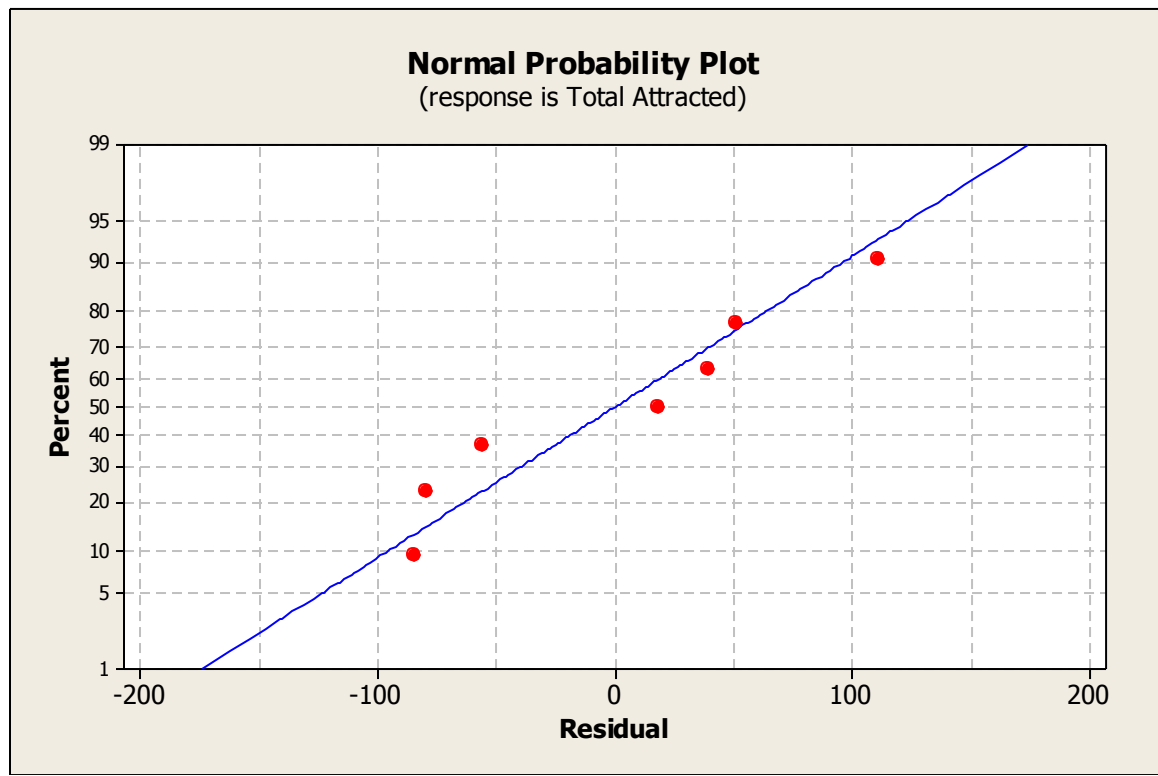
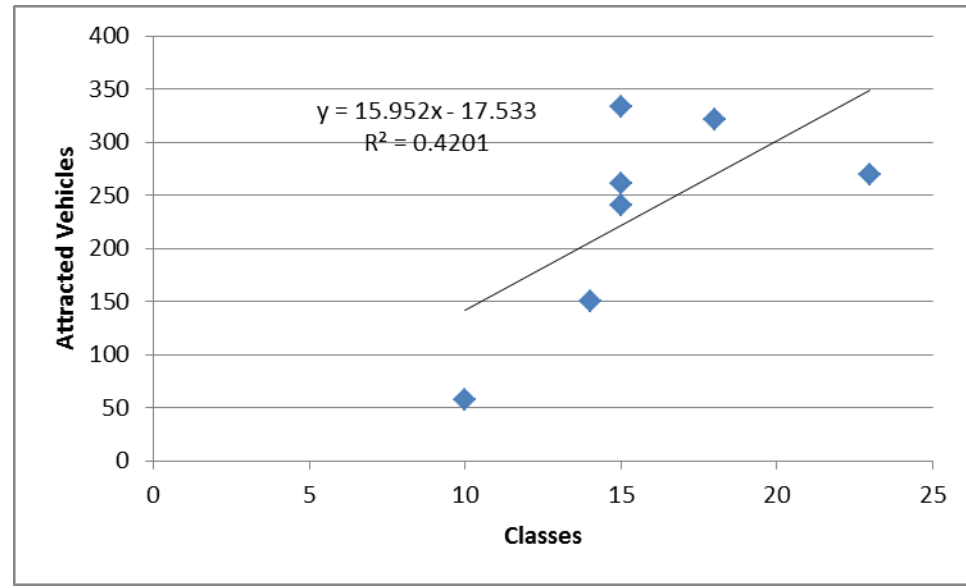
<i>Regression Statistics</i>	
Multiple R	0.984794
R Square	0.969819
Adjusted R Square	0.803152
Standard Error	46.95462
Observations	7

ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	425071.6	425071.6	192.7993	3.481E-05
Residual	6	13228.42	2204.736		
Total	7	438300			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Employees	5.852078634	0.421461025	13.88522	8.69E-06

Classrooms





SUMMARY OUTPUT

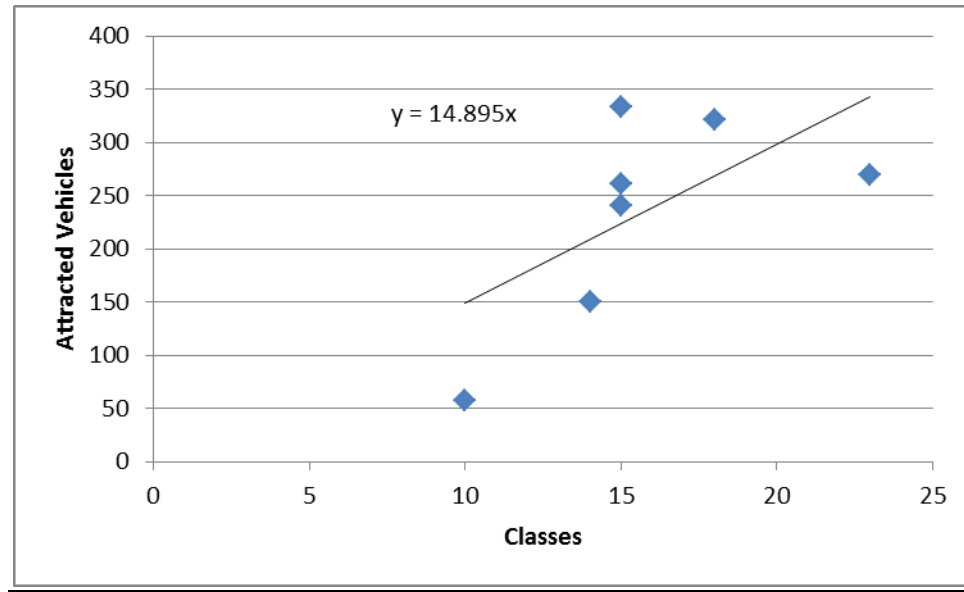
<i>Regression Statistics</i>	
Multiple R	0.648115
R Square	0.420053
Adjusted R Square	0.304064
Standard Error	81.88676
Observations	7

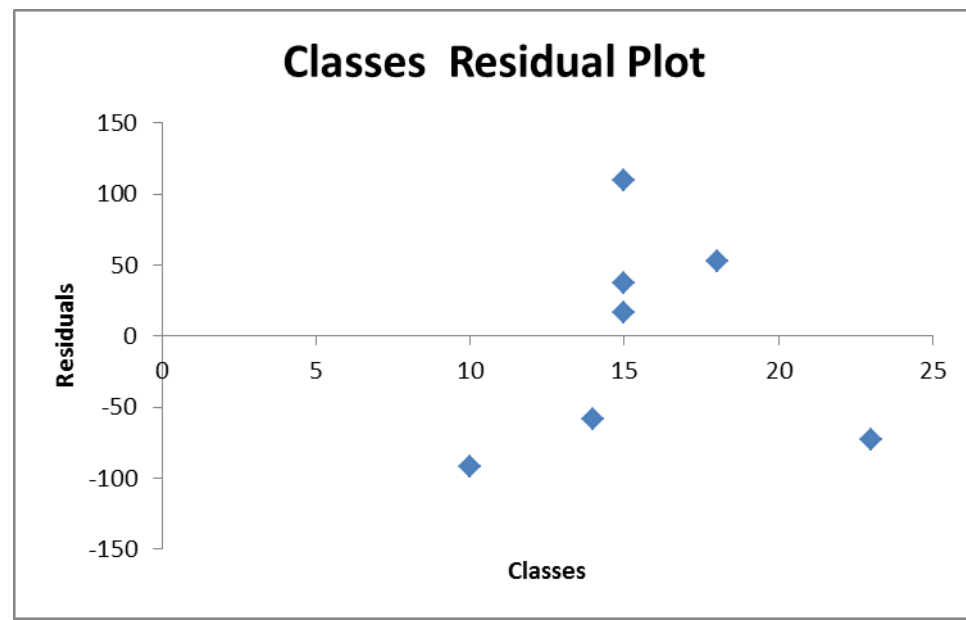
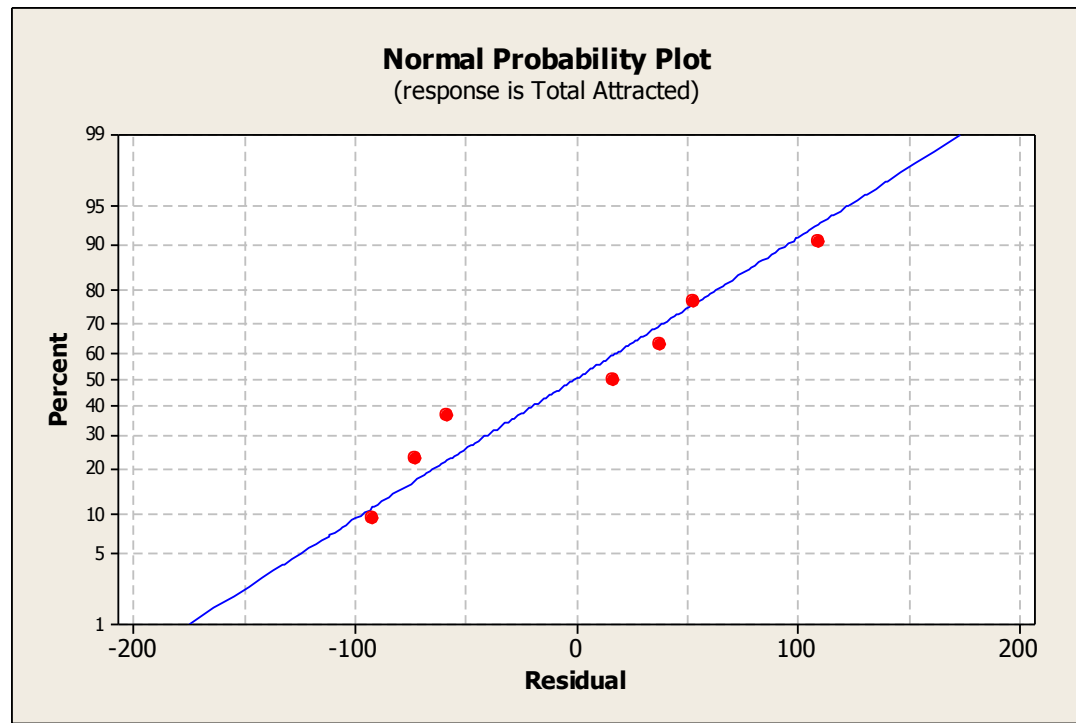
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	24283.65	24283.65	3.621484	0.115415236
Residual	5	33527.21	6705.442		
Total	6	57810.86			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	17.53293413	135.3124907	-0.12957	0.901955
Classes	15.95209581	8.382517934	1.90302	0.115415

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





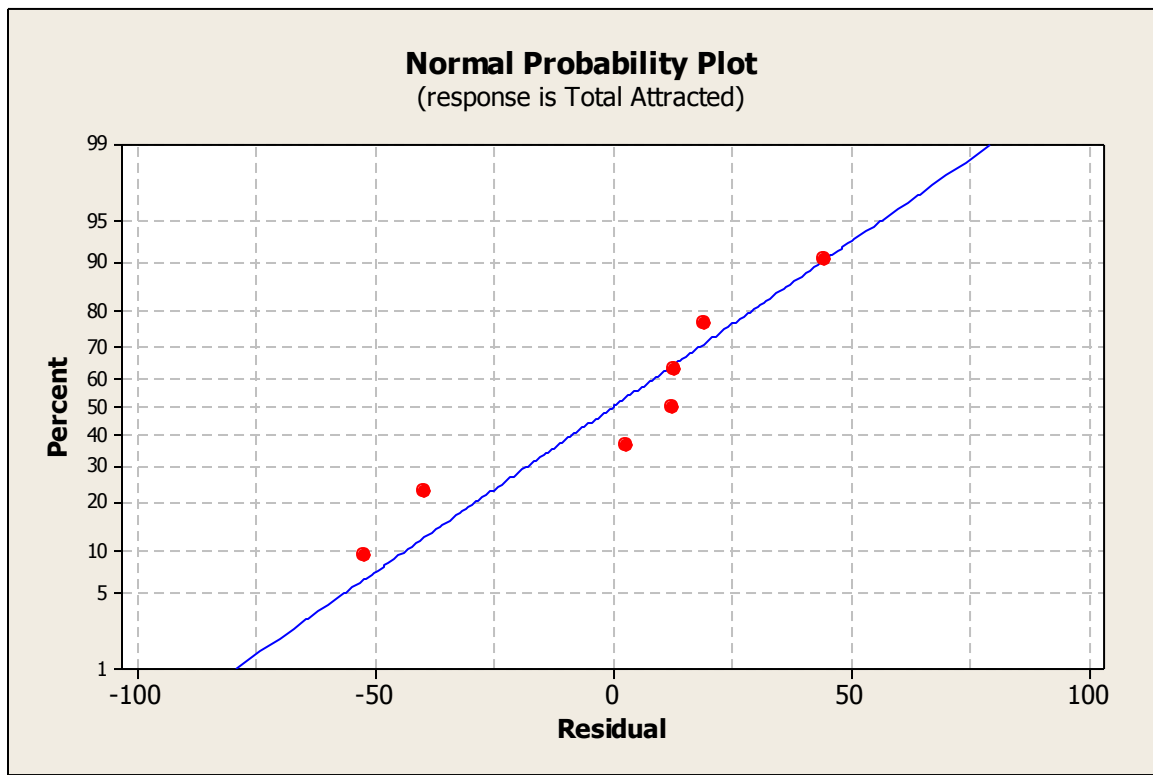
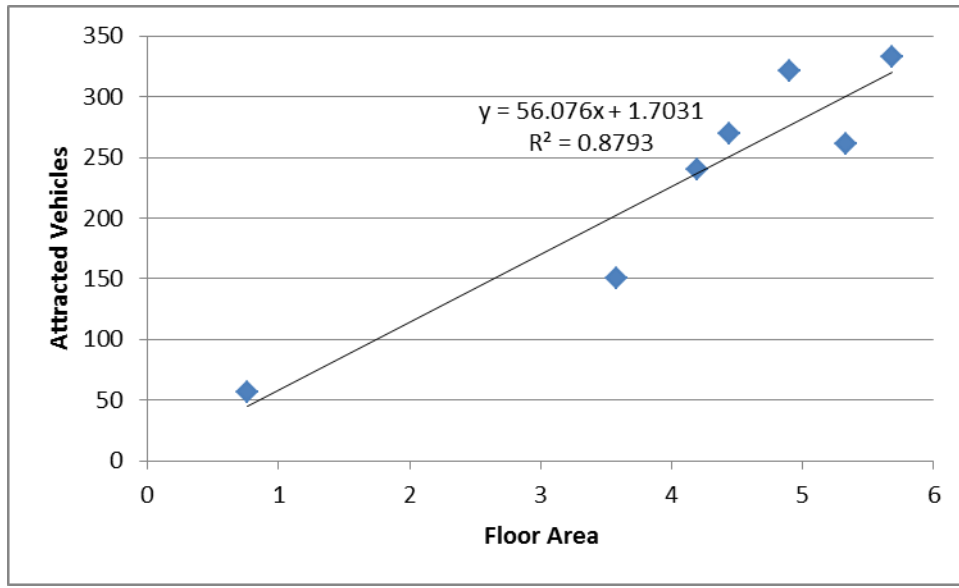
SUMMARY OUTPUT

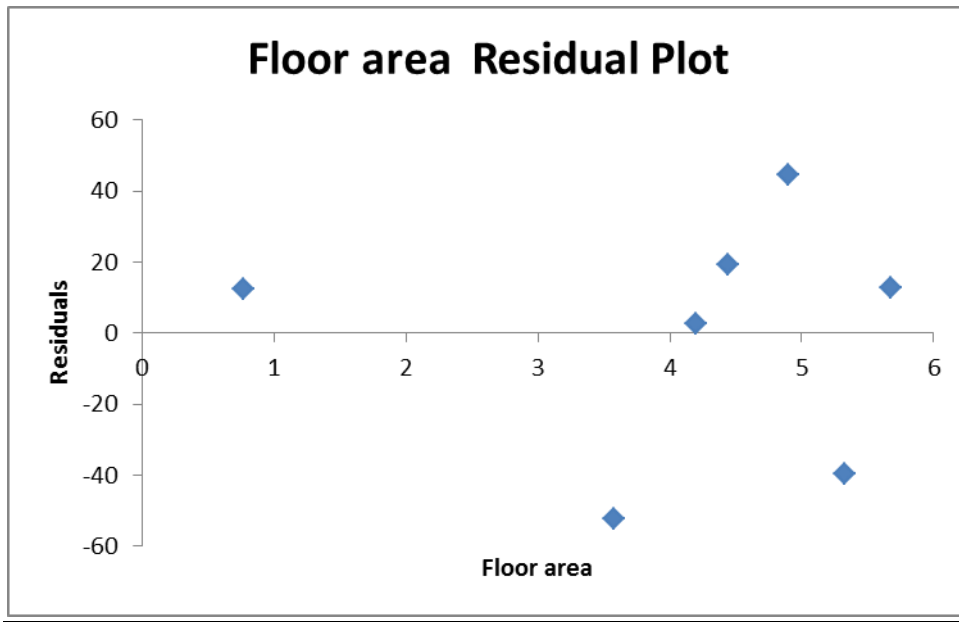
<i>Regression Statistics</i>	
Multiple R	0.960859
R Square	0.923249
Adjusted R Square	0.756583
Standard Error	74.87744
Observations	7

<i>ANOVA</i>					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	404660.2	404660.2	72.17528	0.000371476
Residual	6	33639.79	5606.632		
Total	7	438300			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Classes	14.89473684	1.753228785	8.495604	0.000146

Floor Area





SUMMARY OUTPUT

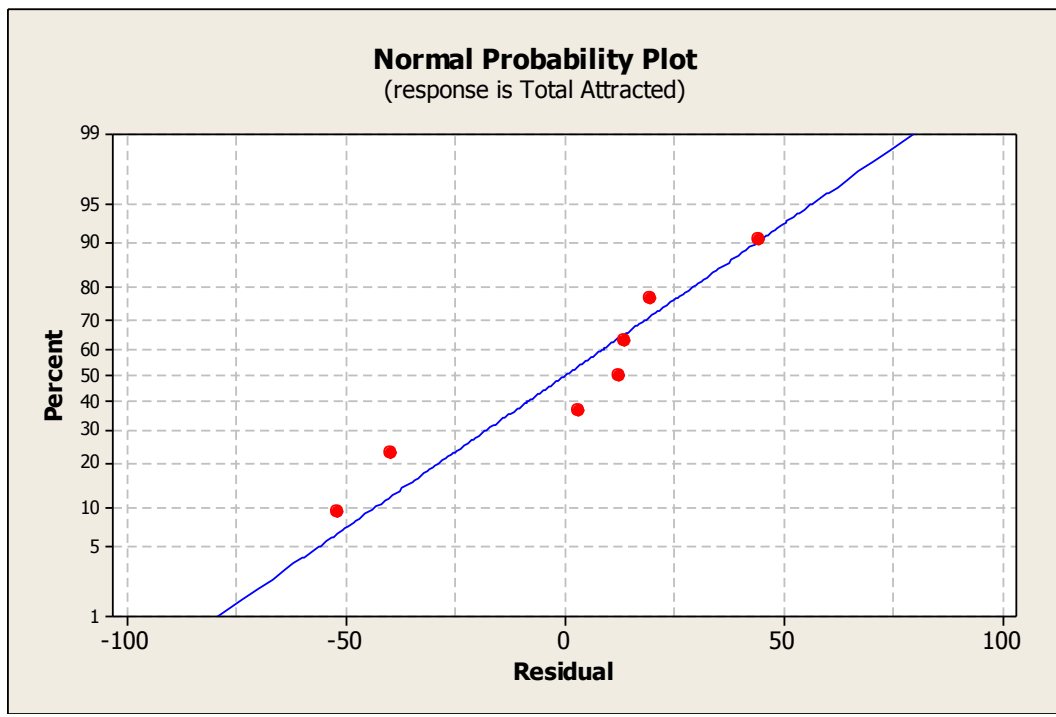
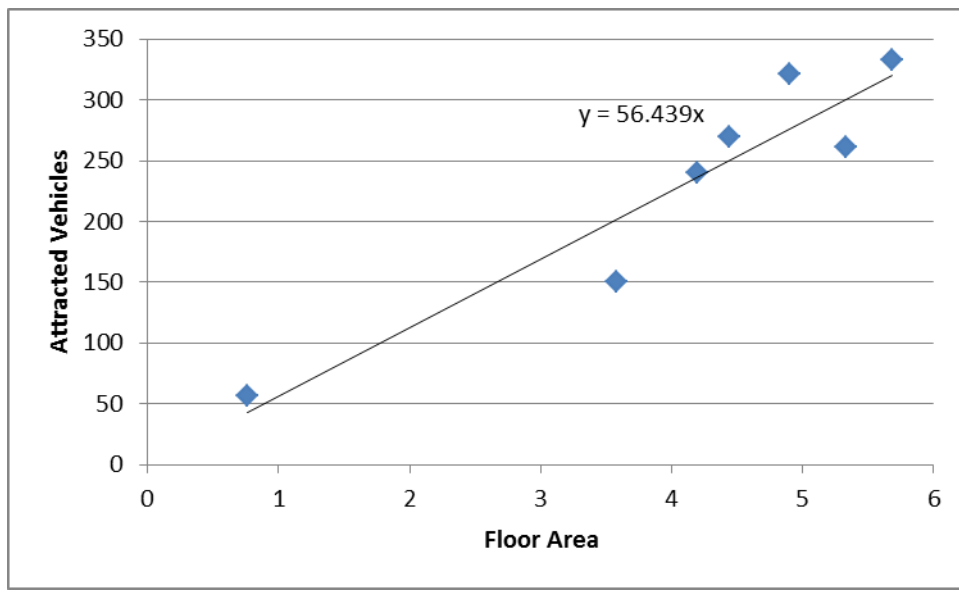
<i>Regression Statistics</i>	
Multiple R	0.937728
R Square	0.879333
Adjusted R Square	0.855199
Standard Error	37.35202
Observations	7

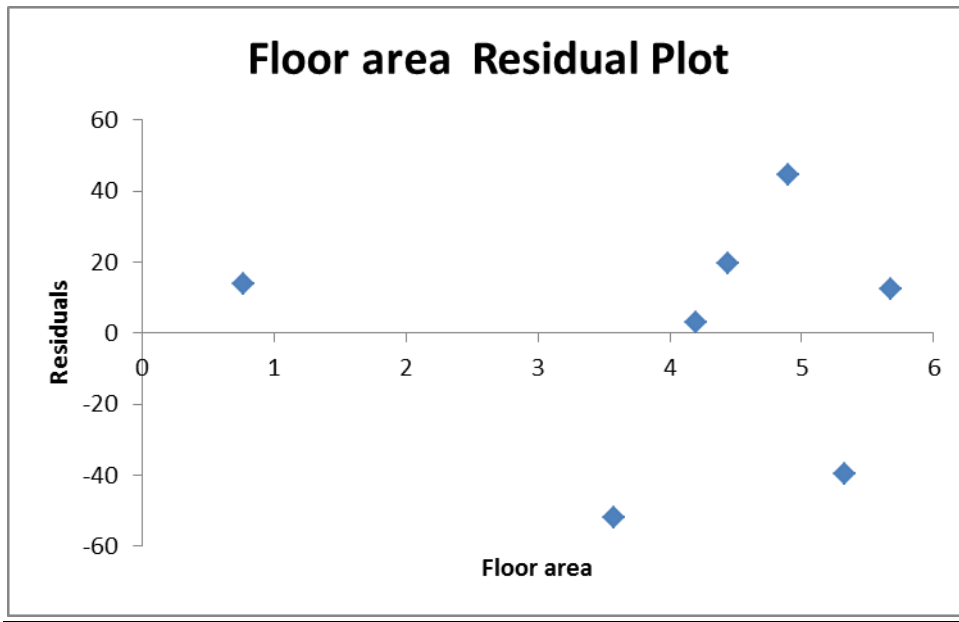
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	50834.99	50834.99	36.43632	0.001797008
Residual	5	6975.868	1395.174		
Total	6	57810.86			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	1.703059086	40.85820076	0.041682	0.968365
Floor area	56.07554555	9.289797666	6.03625	0.001797

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





SUMMARY OUTPUT

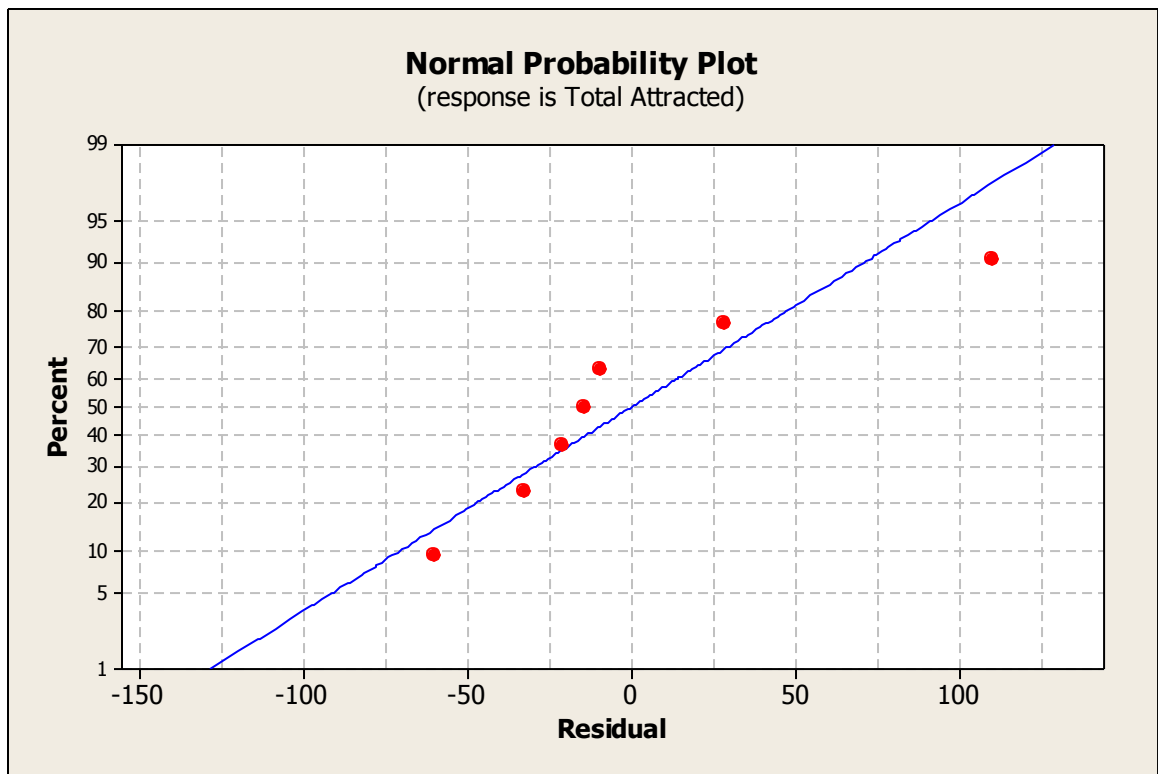
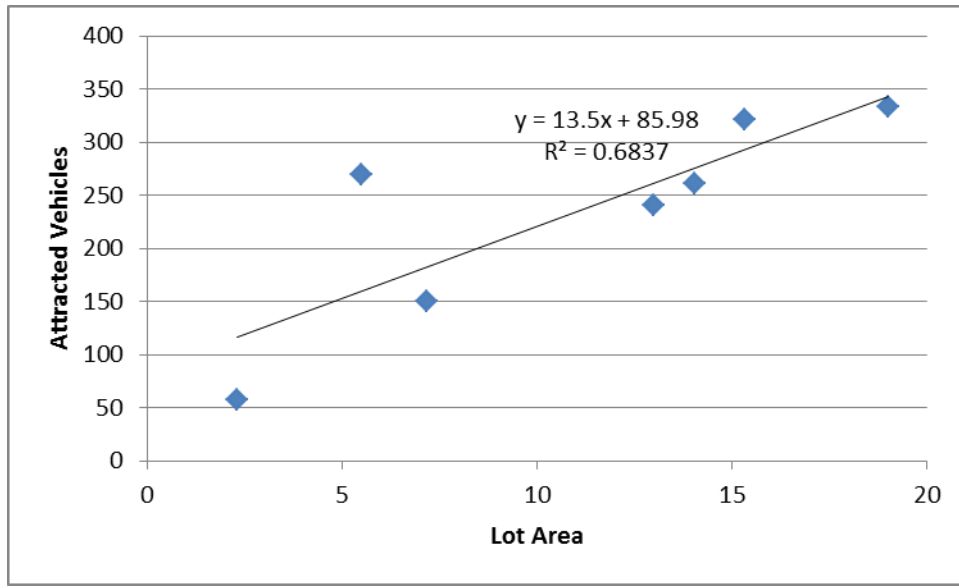
<i>Regression Statistics</i>	
Multiple R	0.992007
R Square	0.984079
Adjusted R Square	0.817412
Standard Error	34.1035
Observations	7

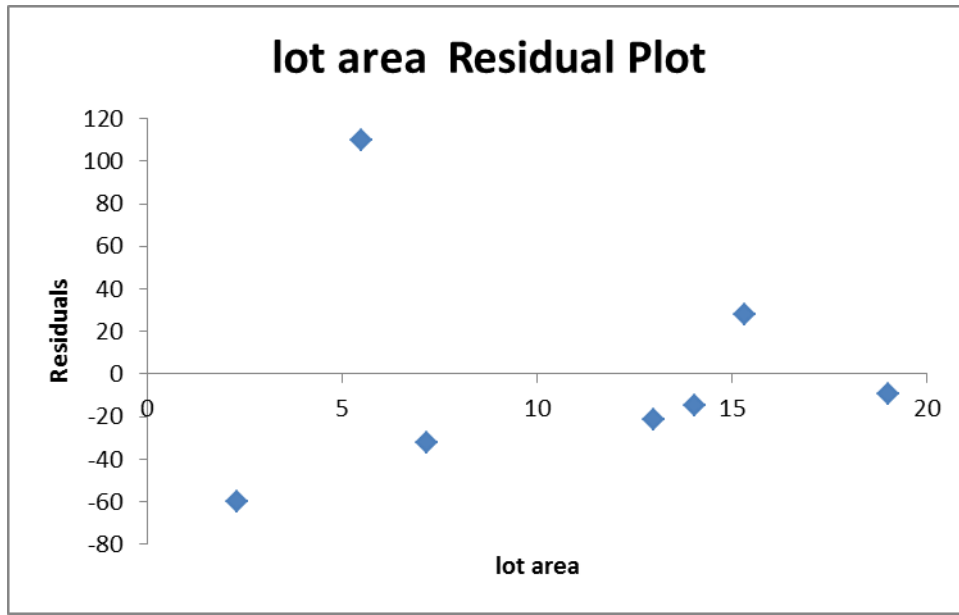
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	431321.7	431321.7	370.8544	6.96353E-06
Residual	6	6978.292	1163.049		
Total	7	438300			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Floor area	56.43891489	2.93073757	19.25758	1.27E-06

Lot Area





SUMMARY OUTPUT

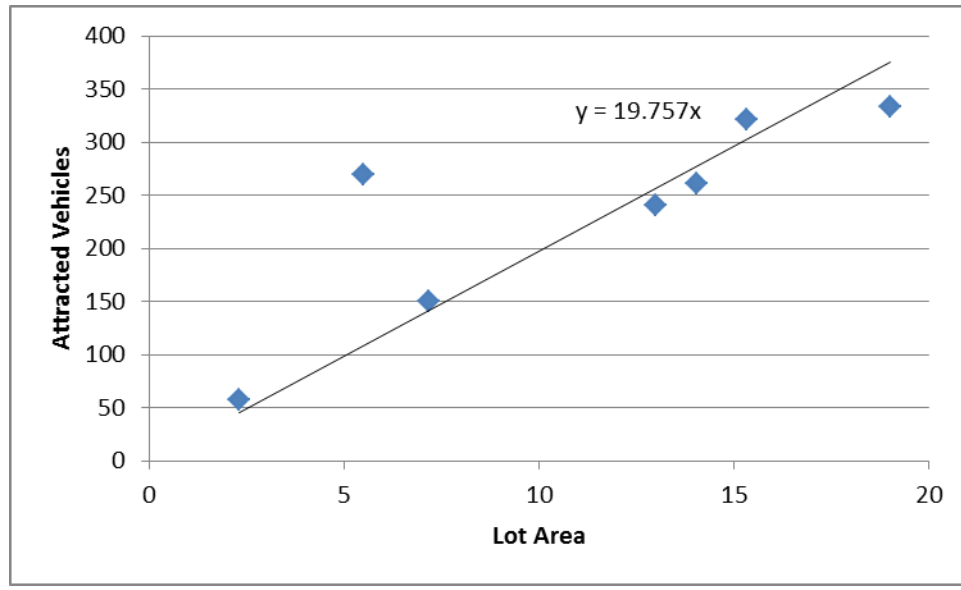
<i>Regression Statistics</i>	
Multiple R	0.826886
R Square	0.683741
Adjusted R Square	0.620489
Standard Error	60.47019
Observations	7

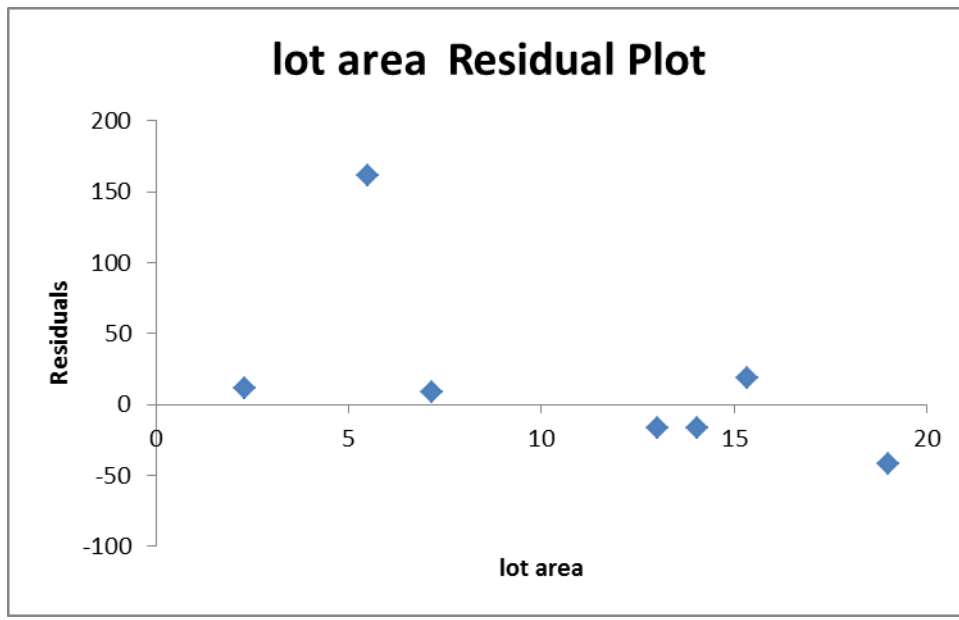
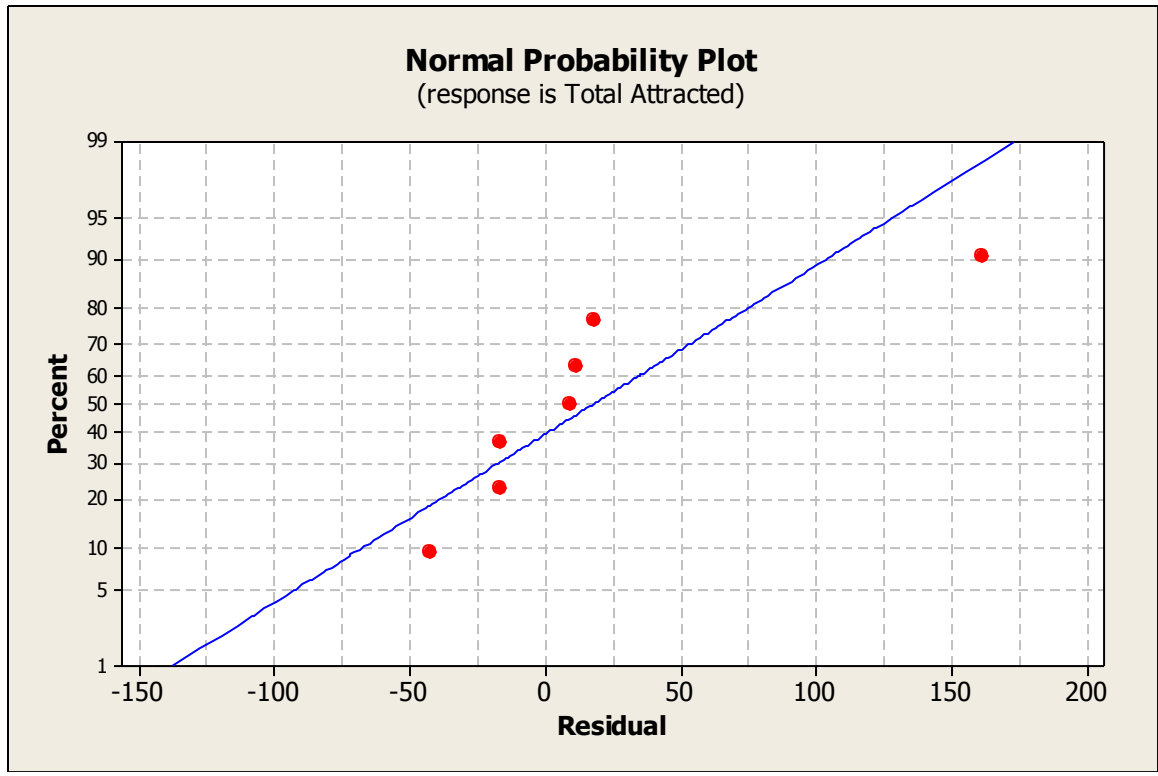
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	39527.64	39527.64	10.80981	0.021765617
Residual	5	18283.22	3656.644		
Total	6	57810.86			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	85.97951002	50.25773775	1.710772	0.14781
lot area	13.50033982	4.106157896	3.287828	0.021766

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





SUMMARY OUTPUT

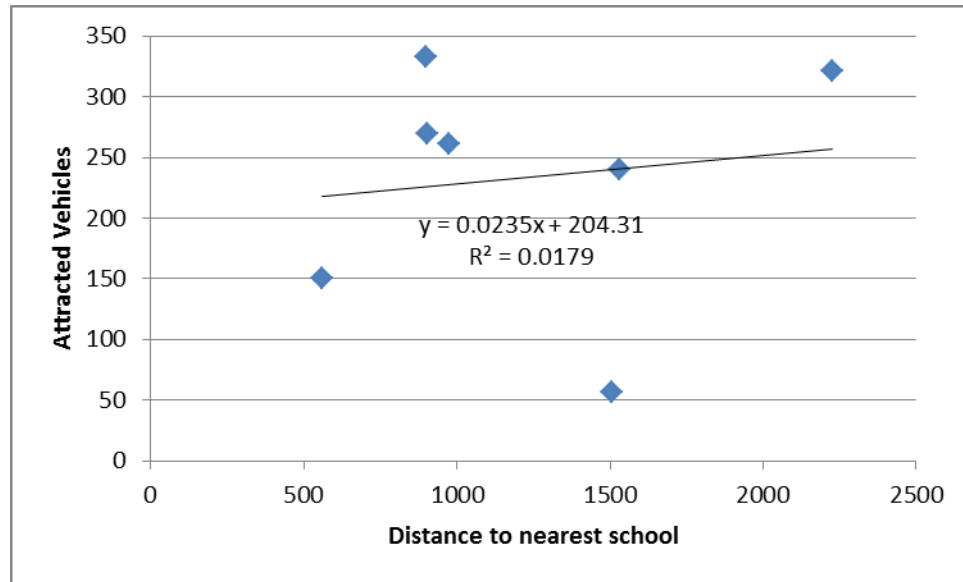
<i>Regression Statistics</i>	
Multiple R	0.966369
R Square	0.933869
Adjusted R Square	0.767202
Standard Error	69.50452
Observations	7

ANOVA

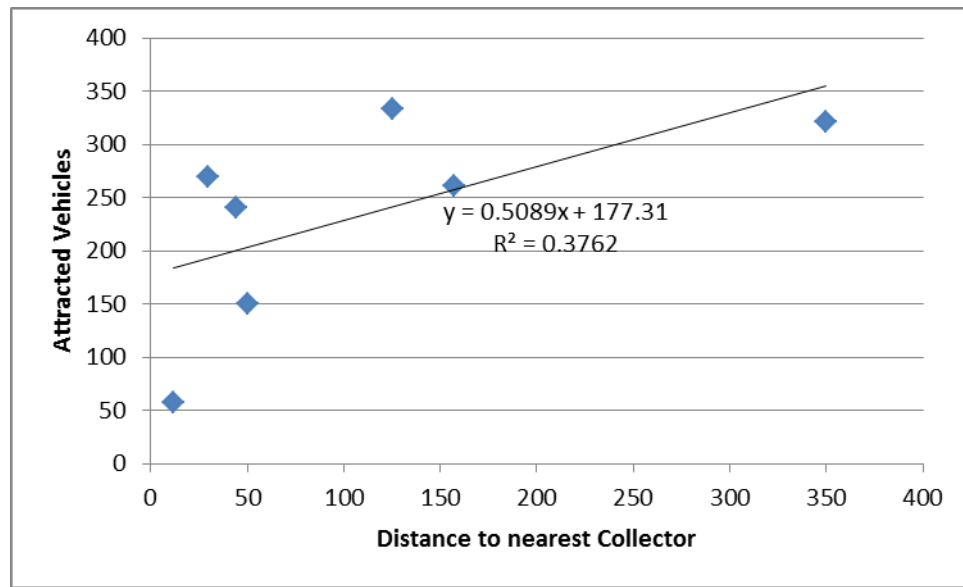
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	409314.7	409314.7	84.72885	0.000253991
Residual	6	28985.27	4830.878		
Total	7	438300			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Lot area	19.75660778	2.146331097	9.204828	9.27E-05

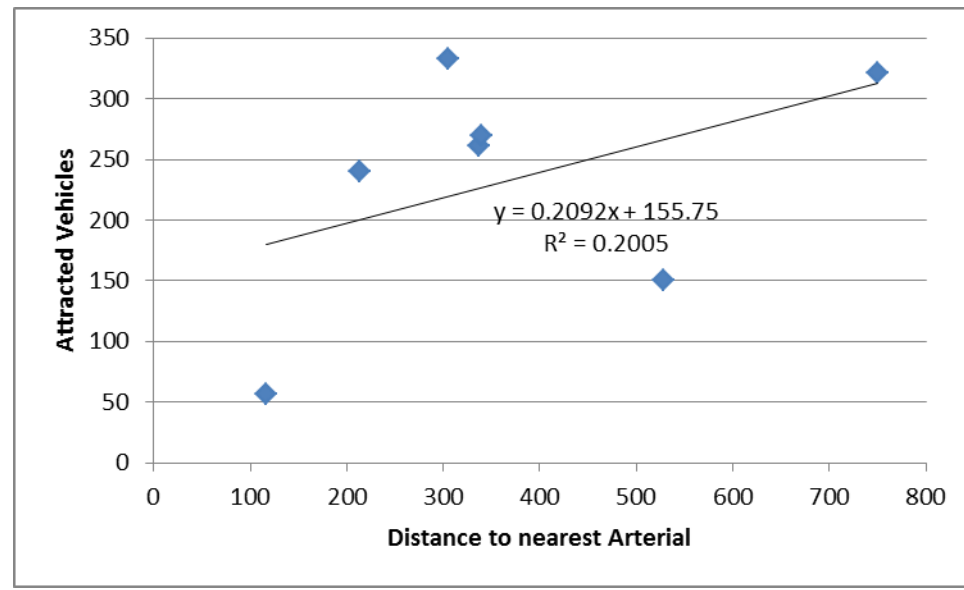
Distance to nearest school



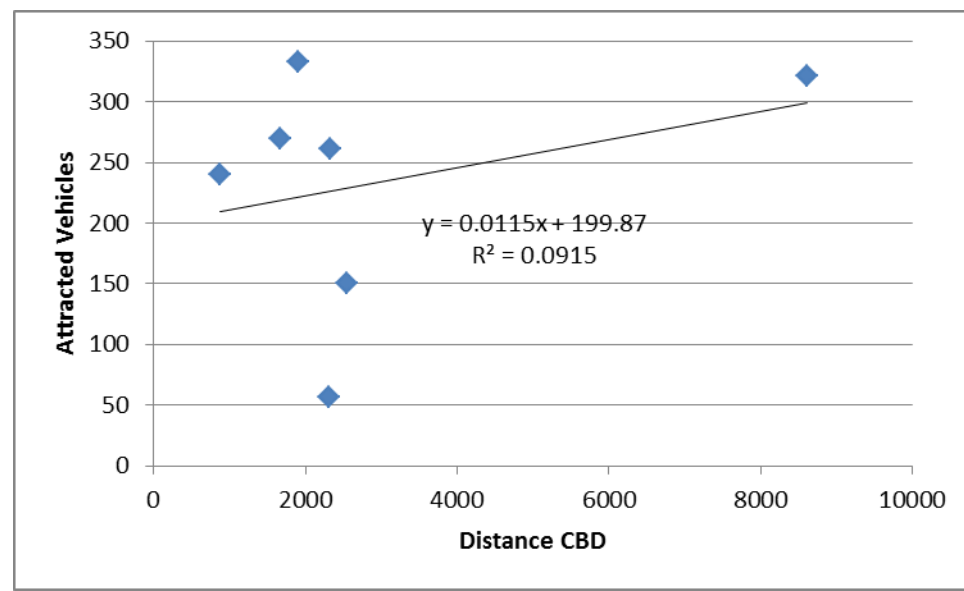
Distance to nearest collector



Distance to the nearest arterial

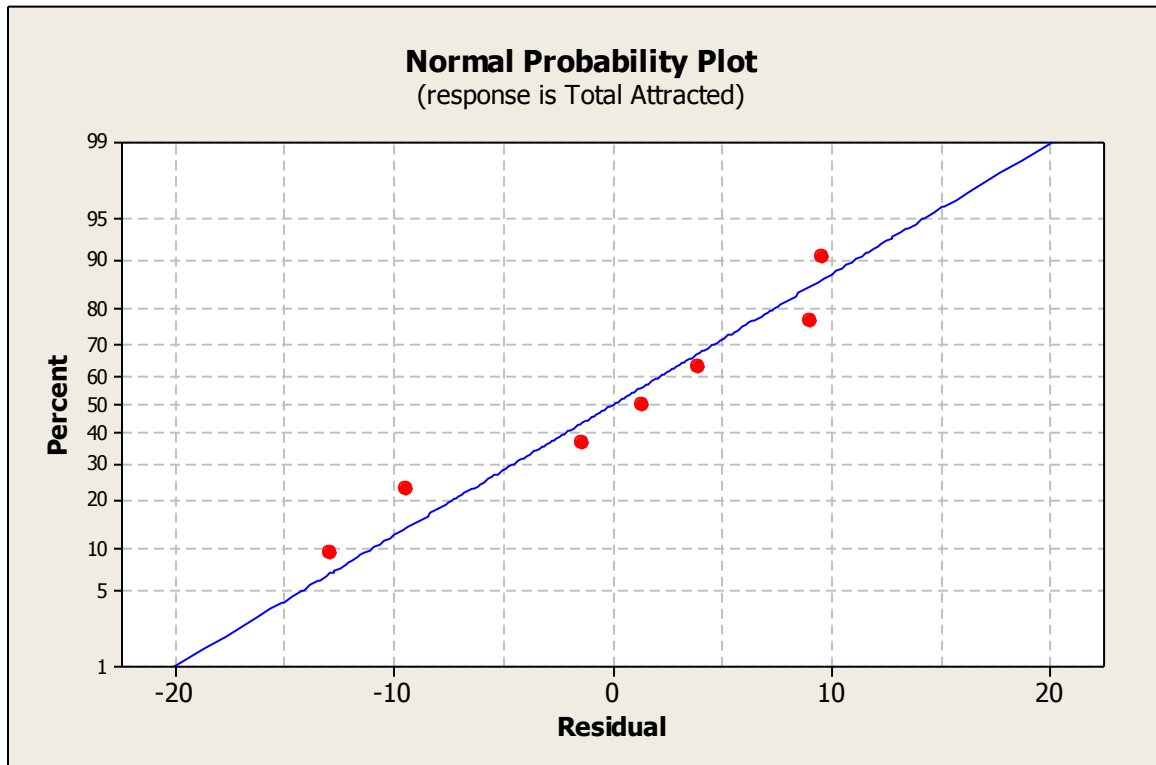


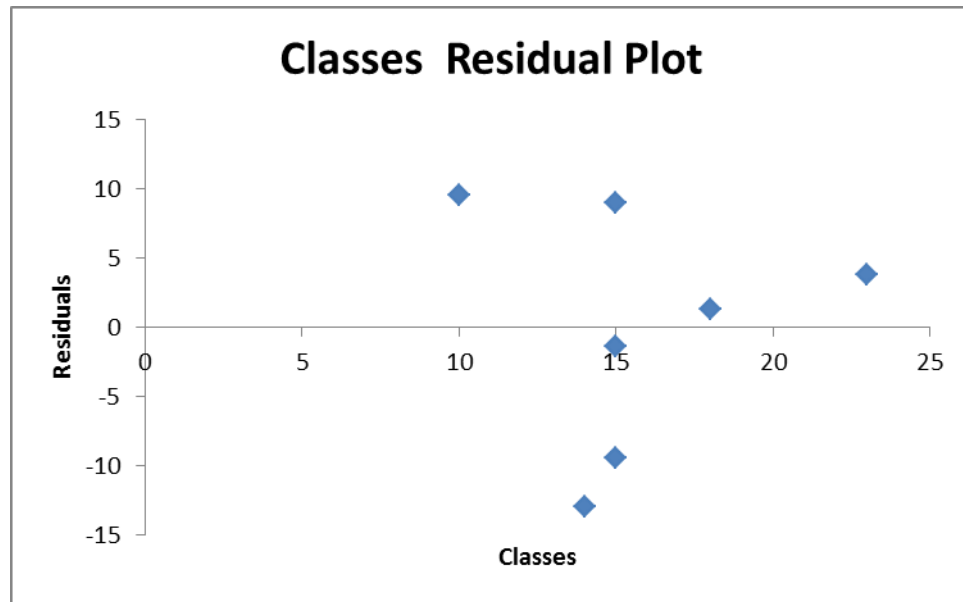
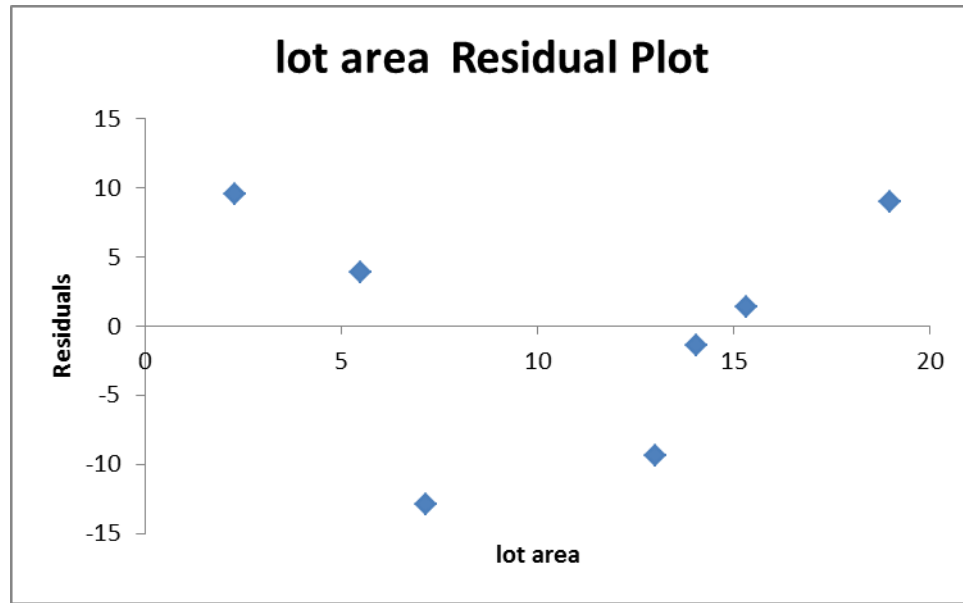
Distance to CBD



Multiple linear regression

Attracted vehicles = $-118.74 + 13.77 * \text{Classes} + 12.44 * \text{Lot Area}$ $R^2 = 0.992$





SUMMARY OUTPUT

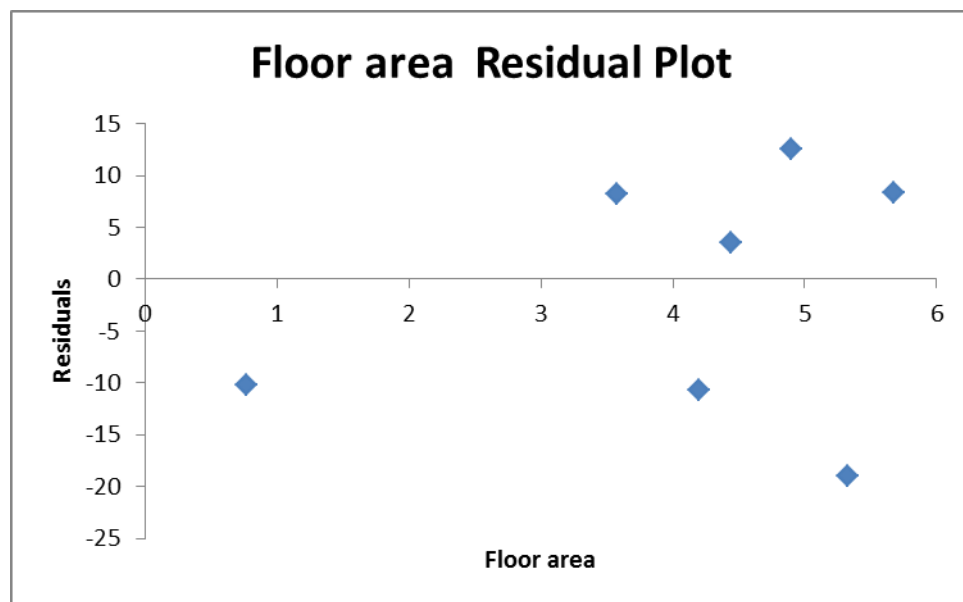
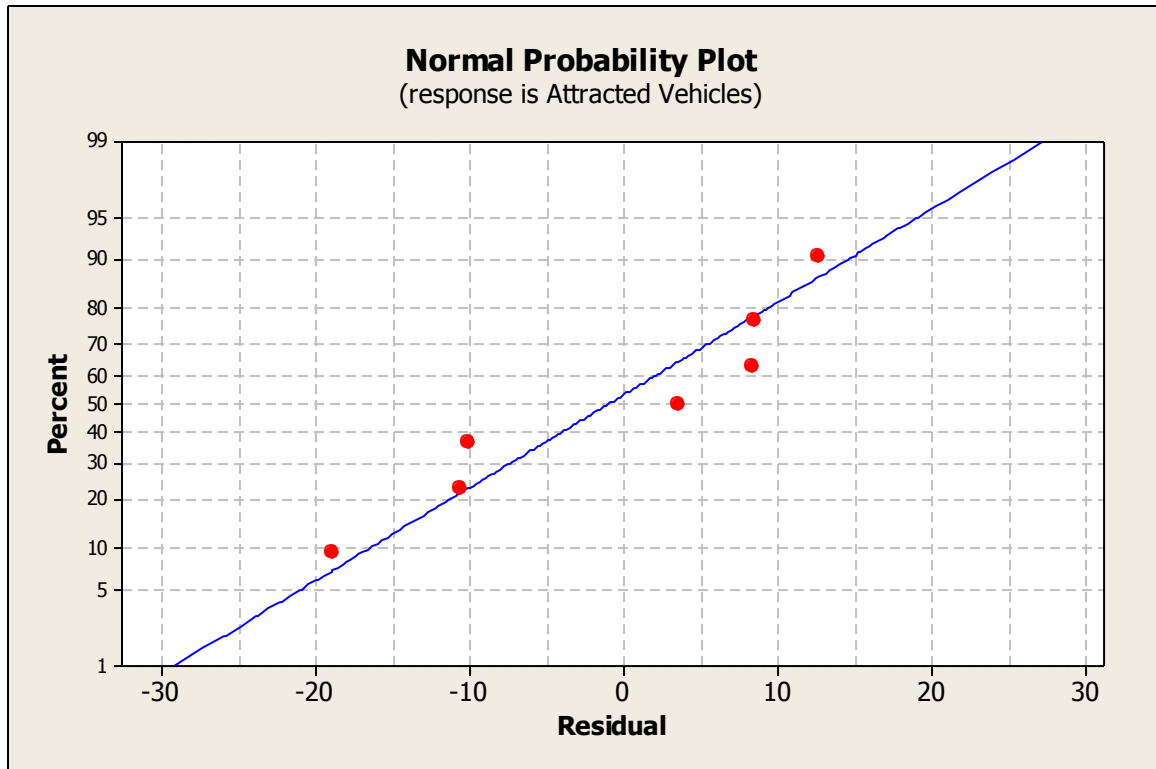
<i>Regression Statistics</i>	
Multiple R	0.996144
R Square	0.992302
Adjusted R Square	0.988453
Standard Error	10.54769
Observations	7

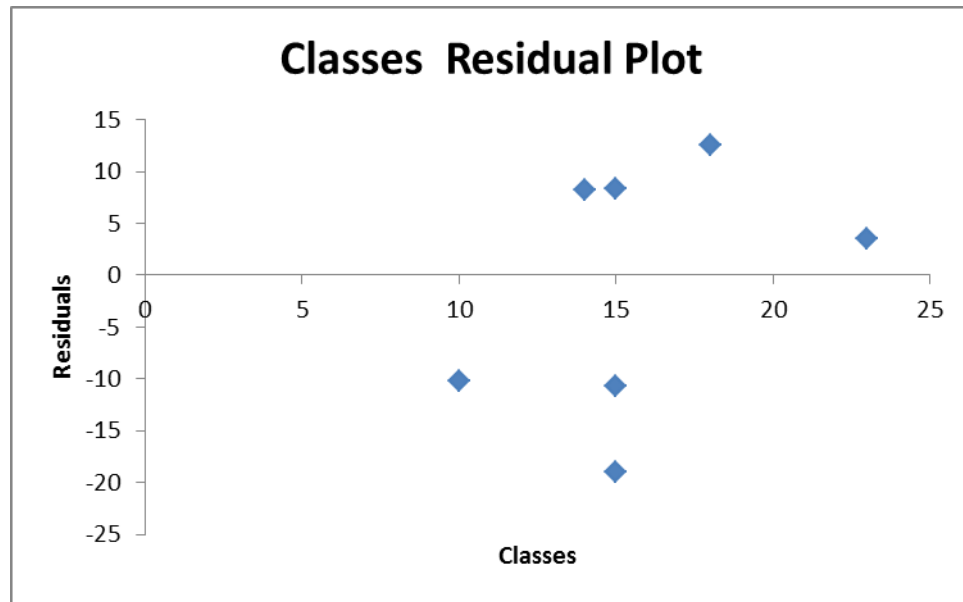
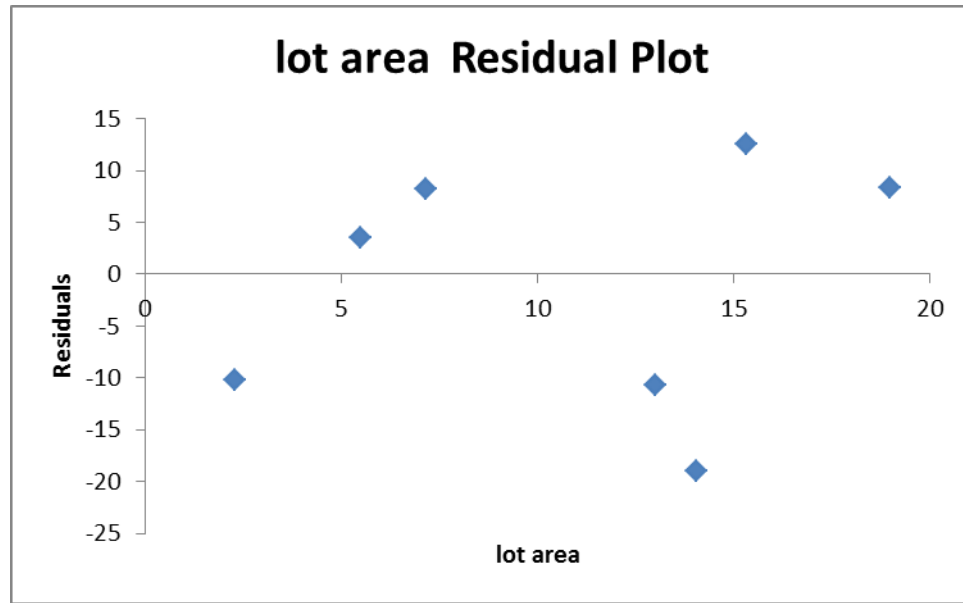
ANOVA

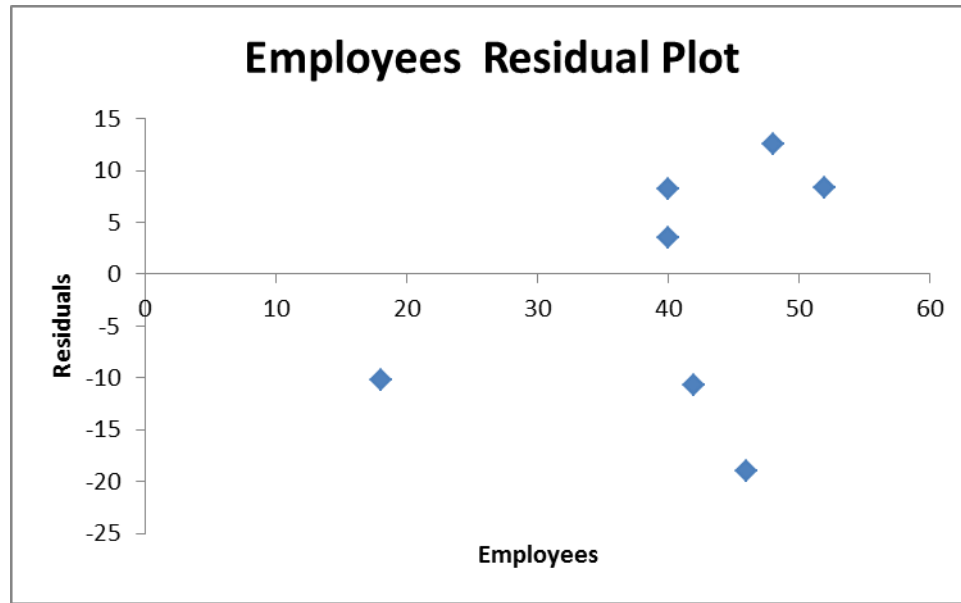
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	2	57365.84	28682.92	257.8155	5.92557E-05
Residual	4	445.0147	111.2537		
Total	6	57810.86			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	-118.737	18.3909388	-6.45627	0.002963
Classes	13.76605	1.087153513	12.66247	0.000224
Lot area	12.43553	0.721147942	17.24408	6.64E-05

$$\text{Attracted vehicles} = 13.049 * \text{Classes} + 14.84 * \text{Lot Area} - 6.98 * \text{Employees} + 36.94 * \text{Floor Area} \quad R^2 = 0.998$$







SUMMARY OUTPUT

<i>Regression Statistics</i>	
Multiple R	0.998985
R Square	0.997972
Adjusted R Square	0.66261
Standard Error	17.2148
Observations	7

ANOVA

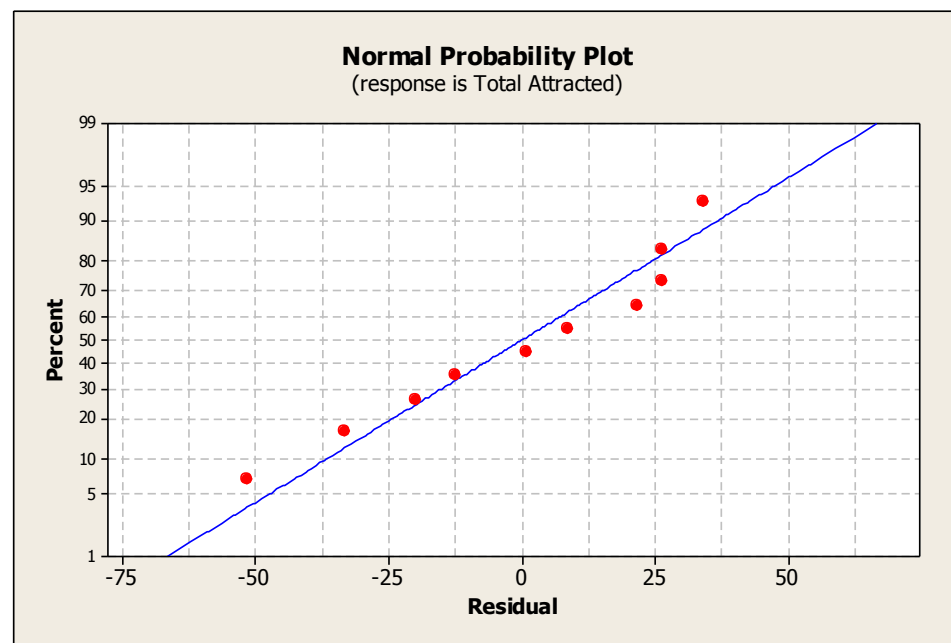
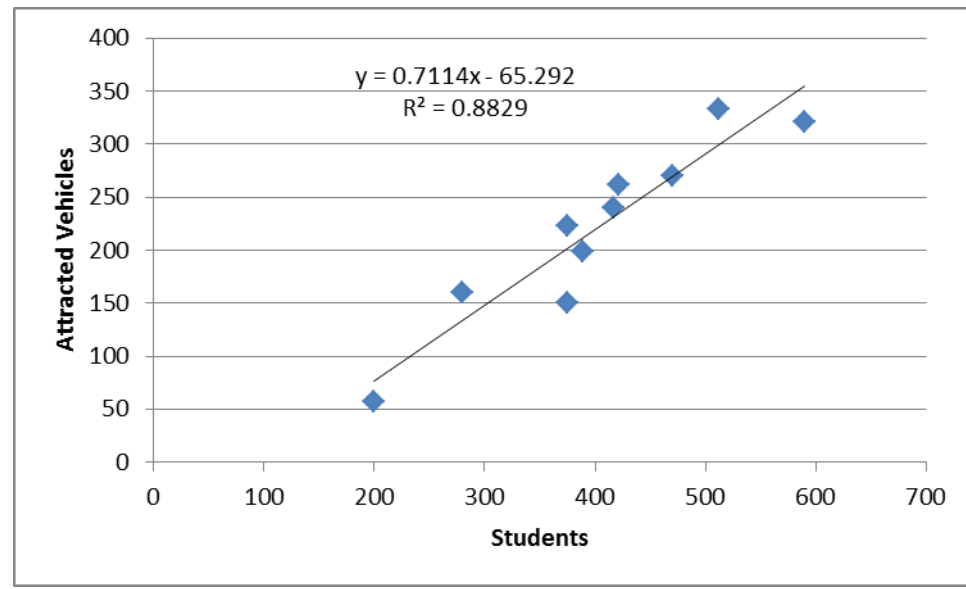
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	4	437411	109352.7	368.9993	0.002704534
Residual	3	889.0483	296.3494		
Total	7	438300			

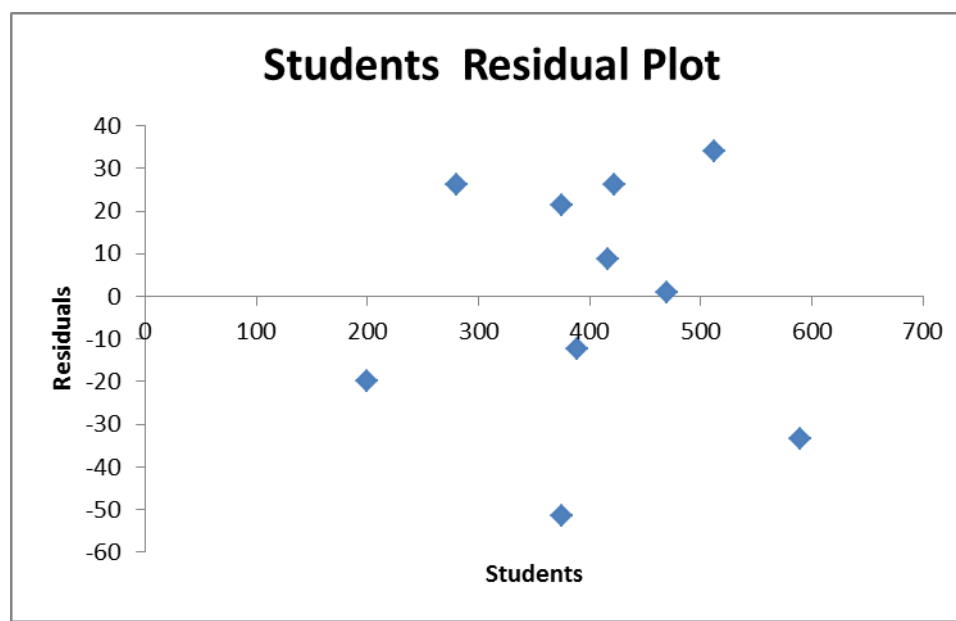
	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Employees	-6.97825868	2.044903997	-3.41251	0.042073
Classes	13.04946904	2.976327232	4.38442	0.021971
Lot area	14.84117211	3.517187294	4.219614	0.02433
Floor area	36.93643246	13.4263769	2.751035	0.070681

Data analysis using all of the 10 schools

Simple linear regression

Students





SUMMARY OUTPUT

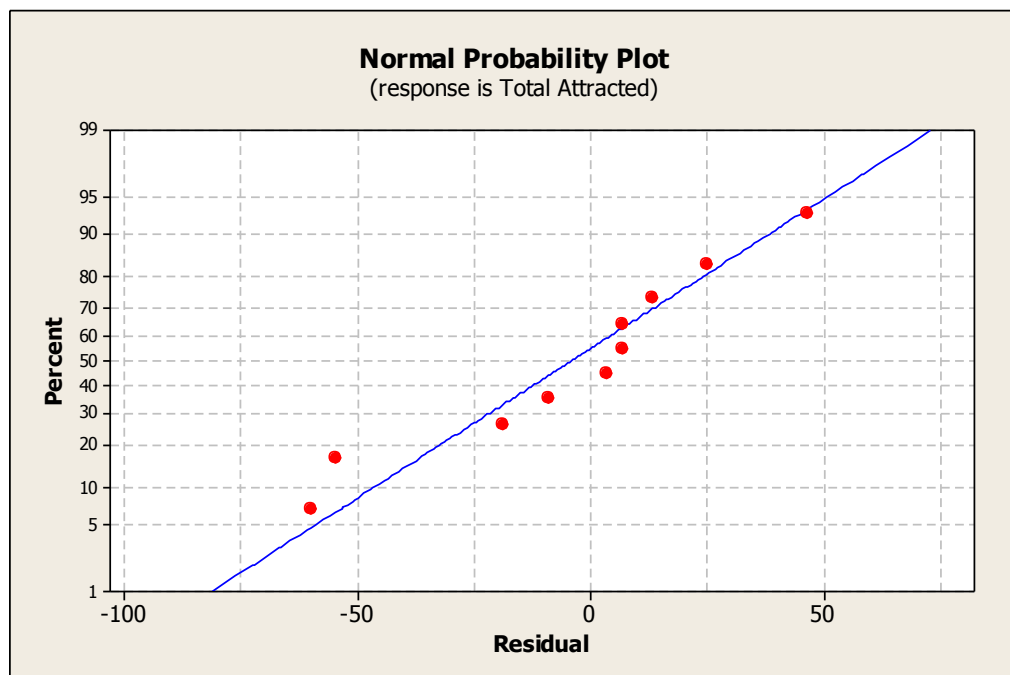
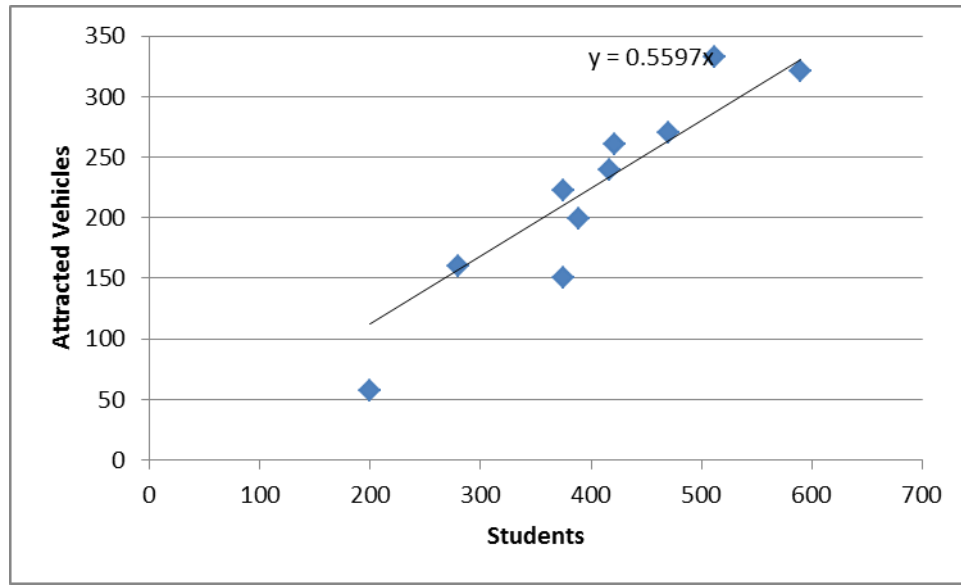
<i>Regression Statistics</i>	
Multiple R	0.939635
R Square	0.882914
Adjusted R Square	0.868278
Standard Error	30.3775
Observations	10

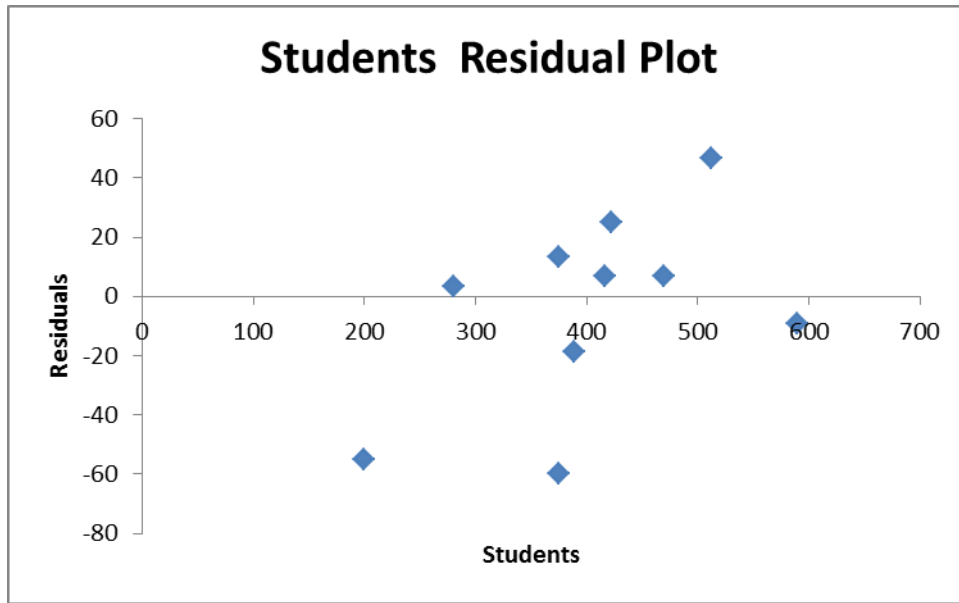
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	55668.06	55668.06	60.32567	5.39896E-05
Residual	8	7382.338	922.7922		
Total	9	63050.4			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	-65.292085	38.14126878	-1.71185	0.125284
Students	0.71139475	0.091592428	7.76696	5.4E-05

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





SUMMARY OUTPUT

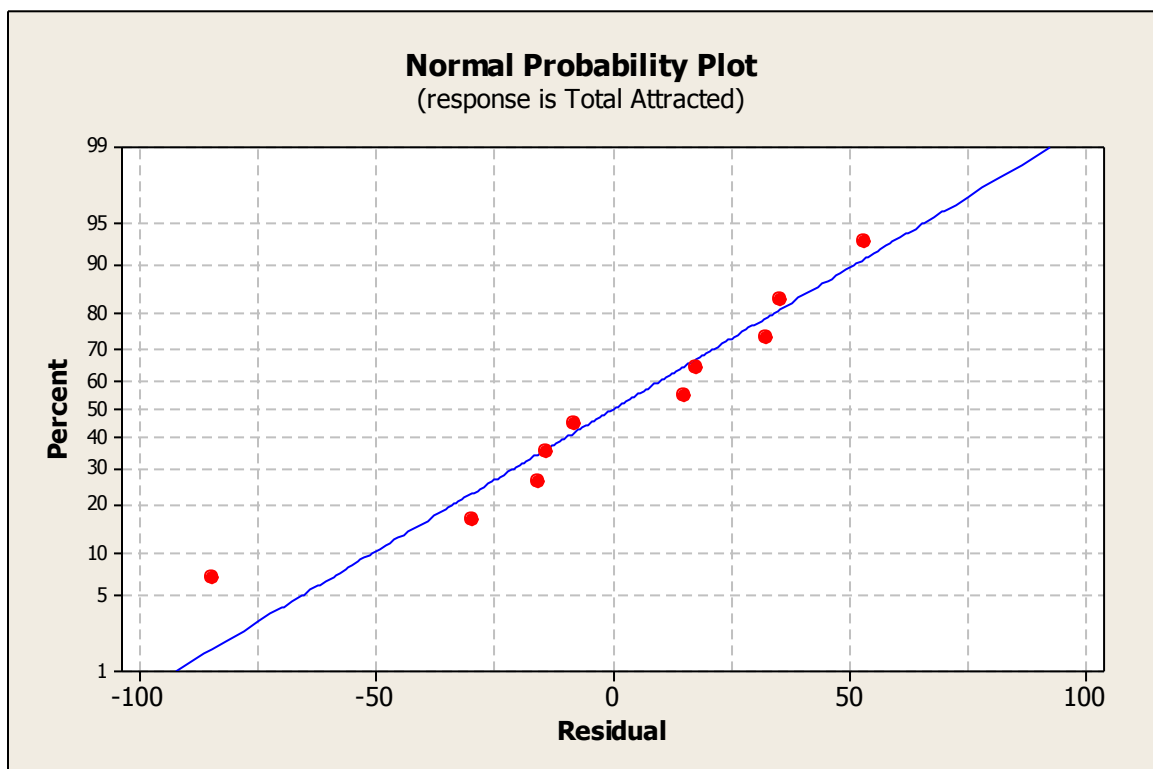
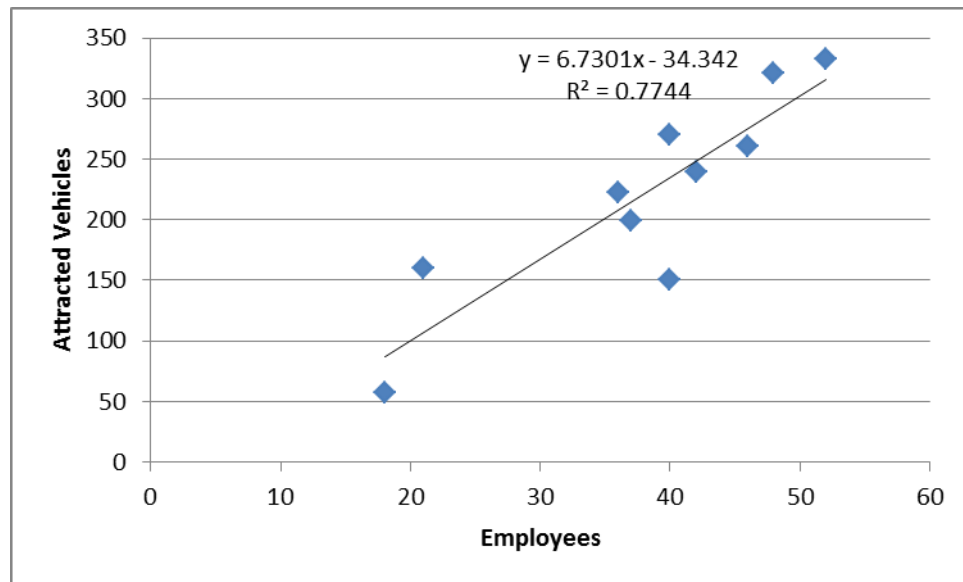
<i>Regression Statistics</i>	
Multiple R	0.990842
R Square	0.981768
Adjusted R Square	0.870657
Standard Error	33.47721
Observations	10

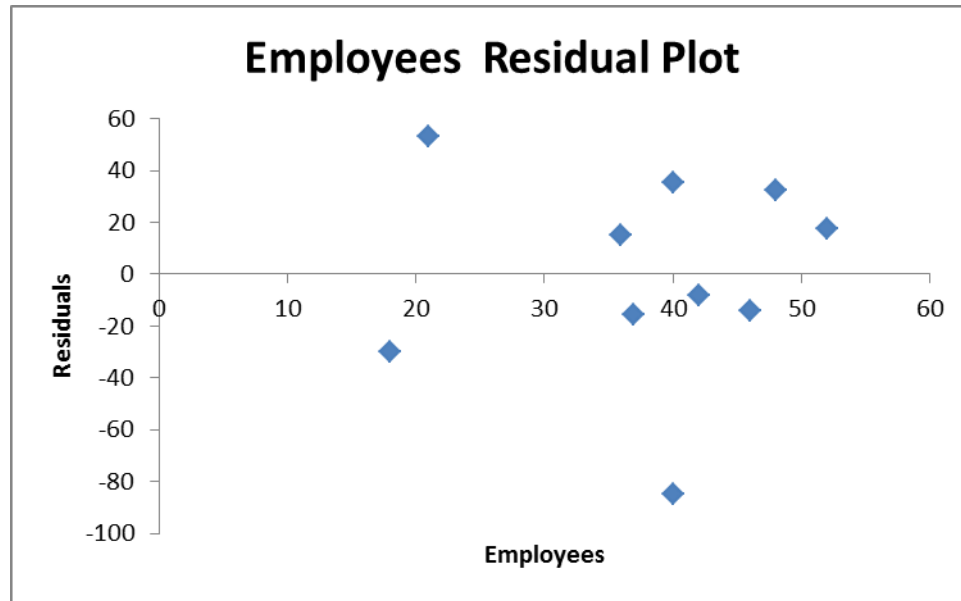
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	543143.5	543143.5	484.6364	1.91405E-08
Residual	9	10086.51	1120.724		
Total	10	553230			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Students	0.559656719	0.025422233	22.01446	3.89E-09

Employees





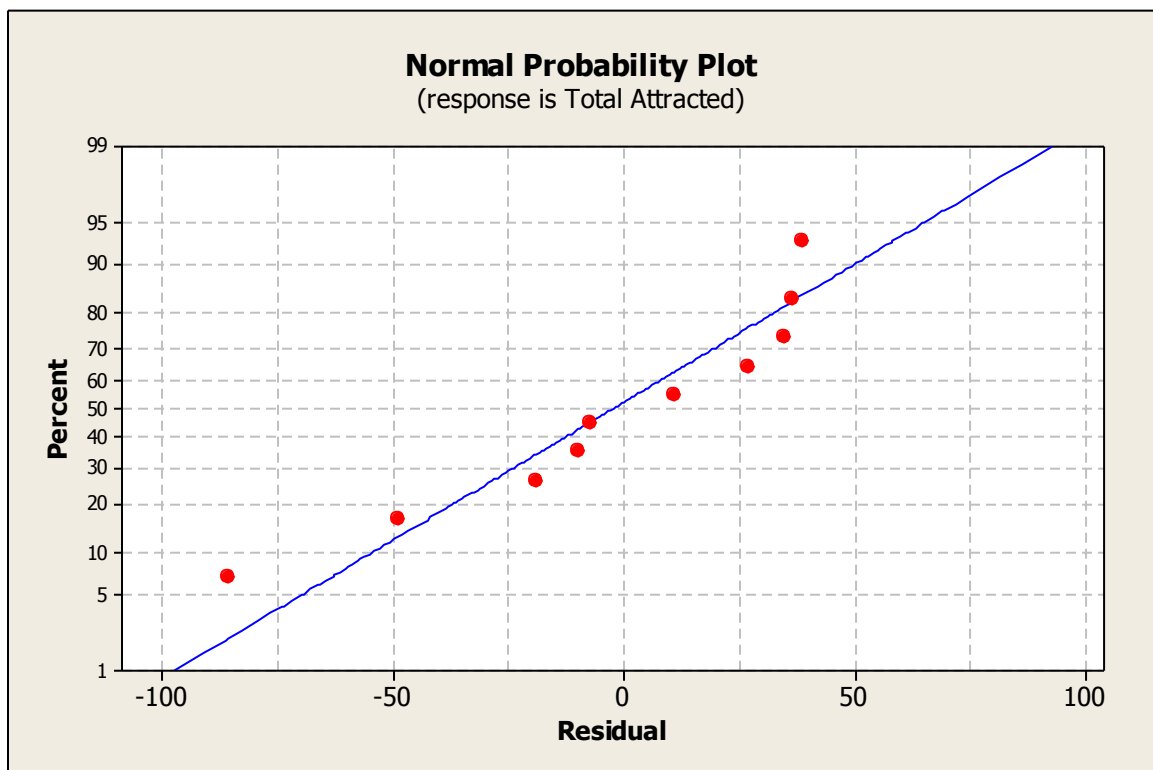
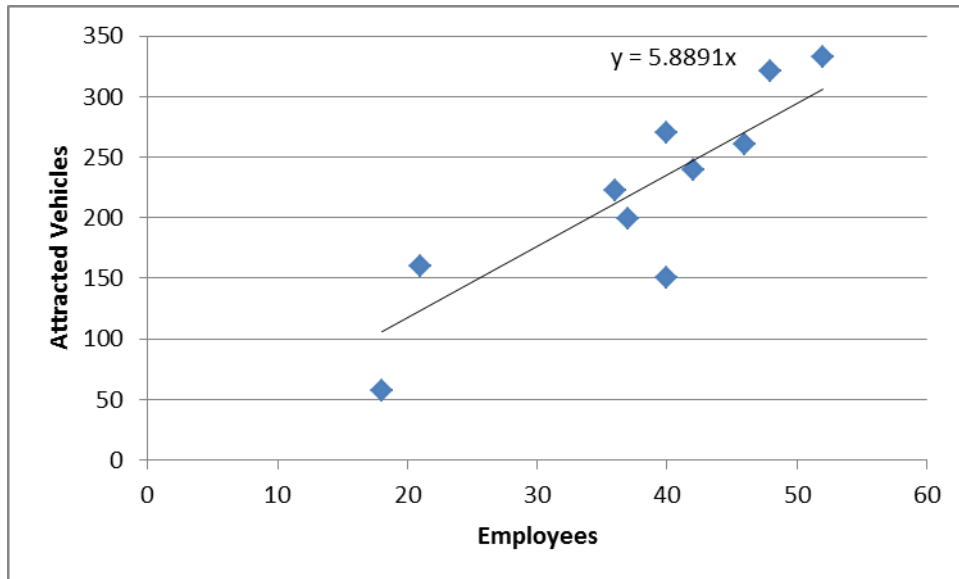
SUMMARY OUTPUT

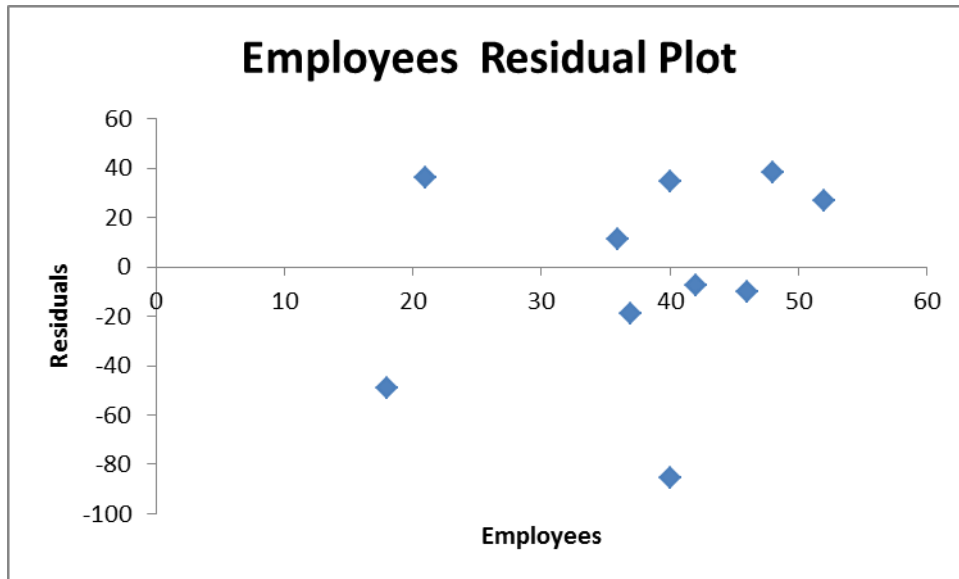
<i>Regression Statistics</i>	
Multiple R	0.880003
R Square	0.774405
Adjusted R Square	0.746206
Standard Error	42.16611
Observations	10

<i>ANOVA</i>					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	48826.55	48826.55	27.4618	0.00078291
Residual	8	14223.85	1777.981		
Total	9	63050.4			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	-34.34211503	50.59085552	-0.67882	0.516405
Employees	6.730055659	1.28426353	5.240401	0.000783

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





SUMMARY OUTPUT

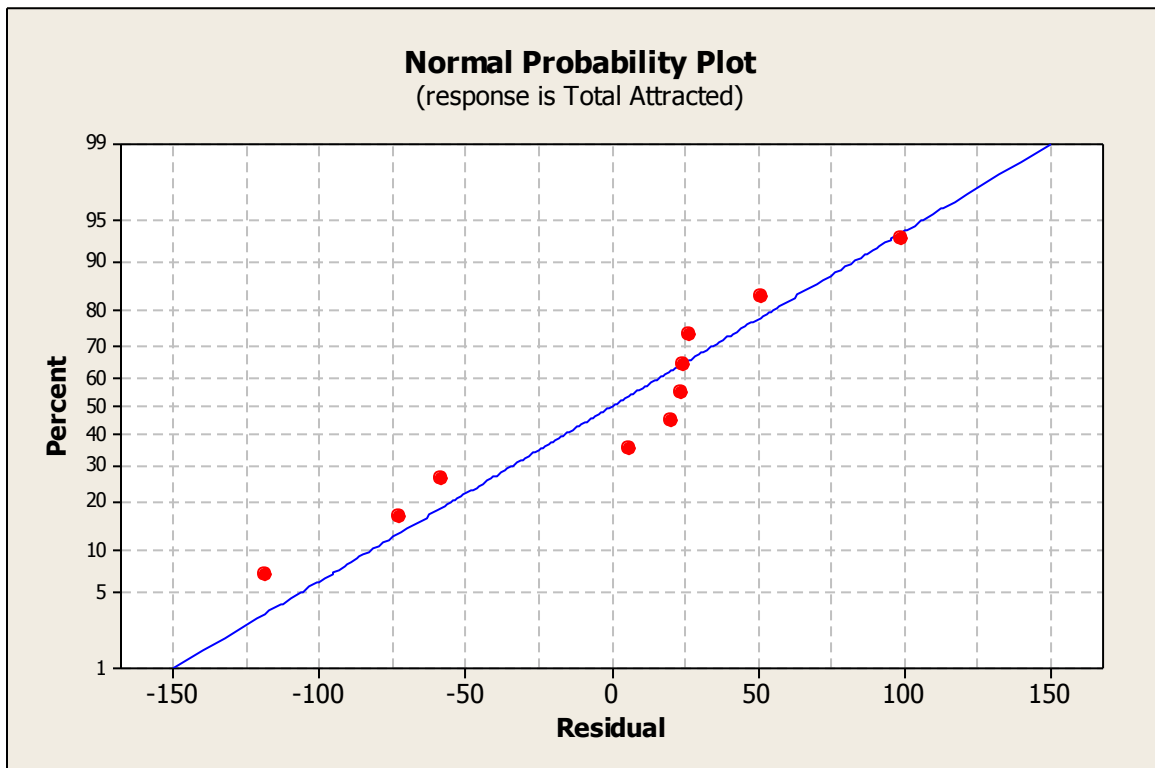
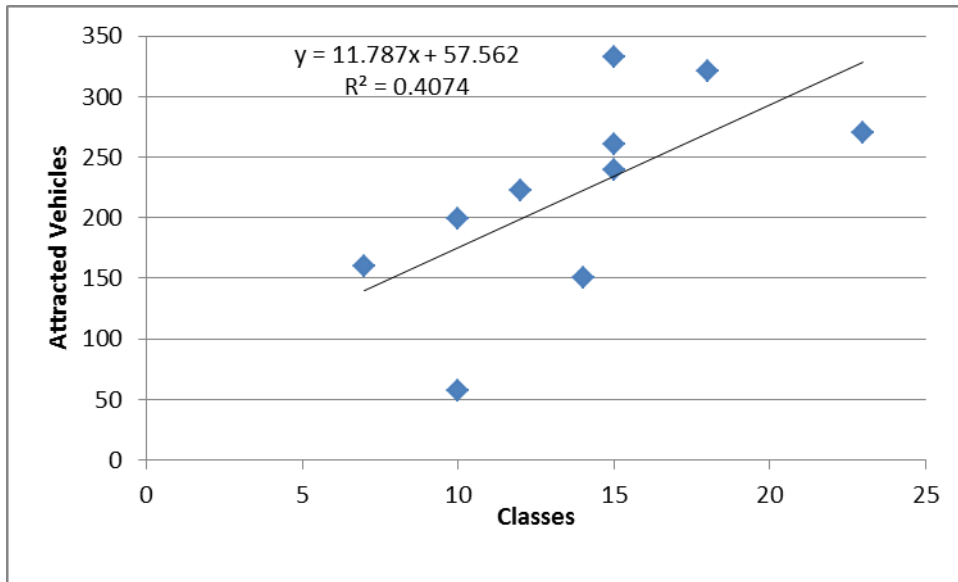
<i>Regression Statistics</i>	
Multiple R	0.986311
R Square	0.972809
Adjusted R Square	0.861697
Standard Error	40.88349
Observations	10

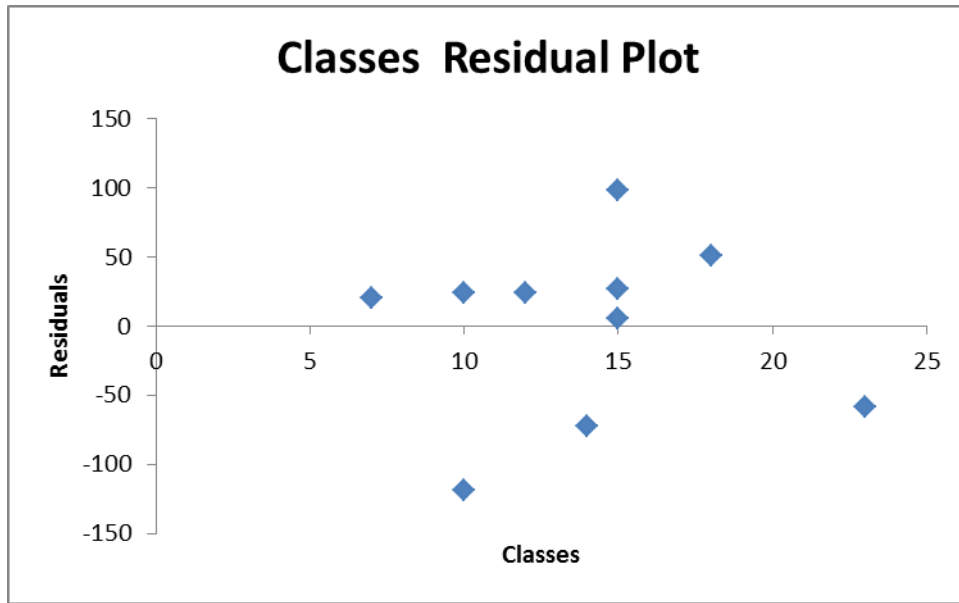
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	538186.9	538186.9	321.9862	9.53873E-08
Residual	9	15043.14	1671.459		
Total	10	553230			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Employees	5.889096533	0.328193556	17.94397	2.36E-08

Classes





SUMMARY OUTPUT

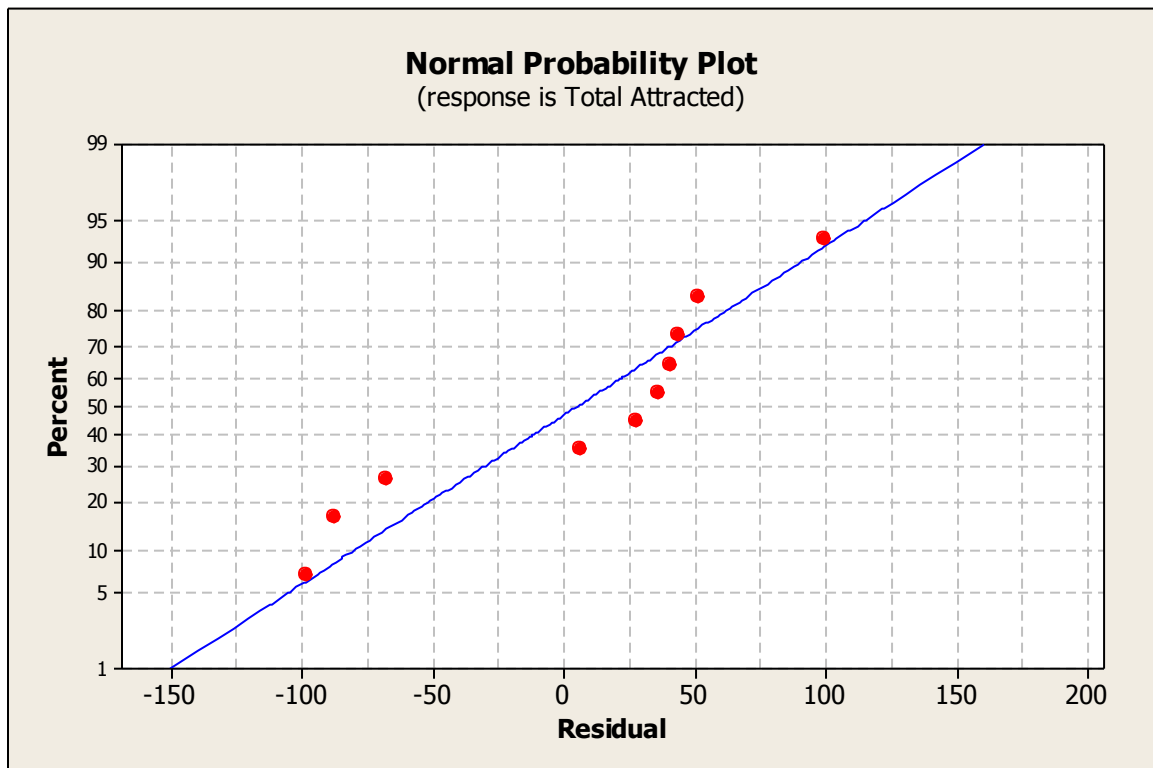
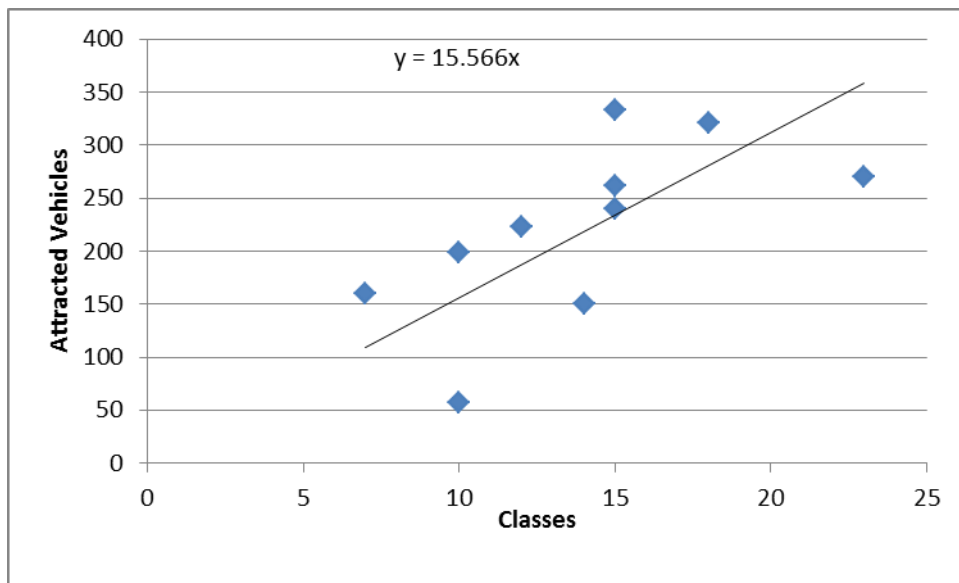
<i>Regression Statistics</i>	
Multiple R	0.6383
R Square	0.407426
Adjusted R Square	0.333355
Standard Error	68.33923
Observations	10

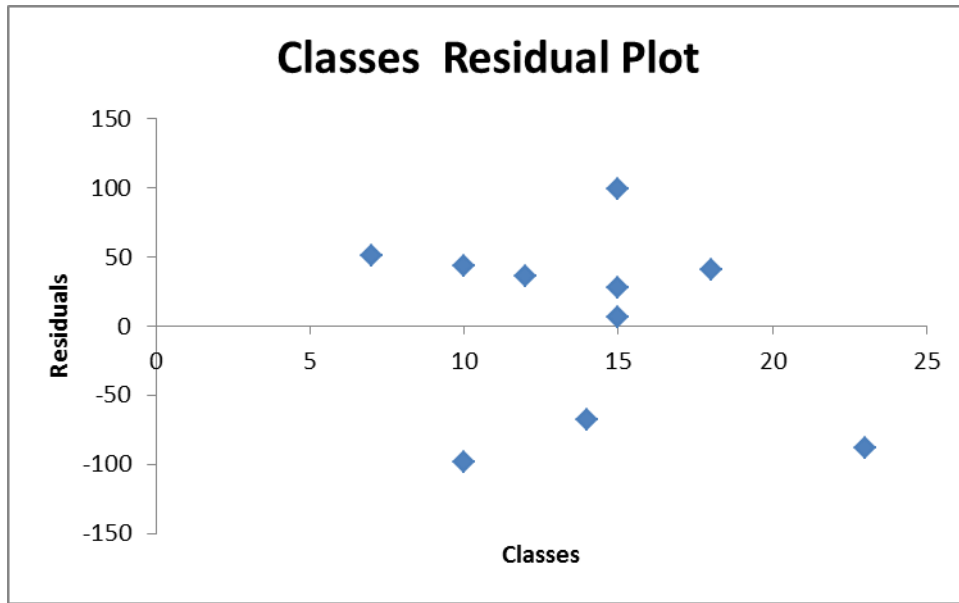
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	25688.4	25688.4	5.500432	0.047025011
Residual	8	37362	4670.251		
Total	9	63050.4			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	57.56192537	73.12435749	0.787179	0.453864
Classes	11.78691184	5.025758712	2.3453	0.047025

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





SUMMARY OUTPUT

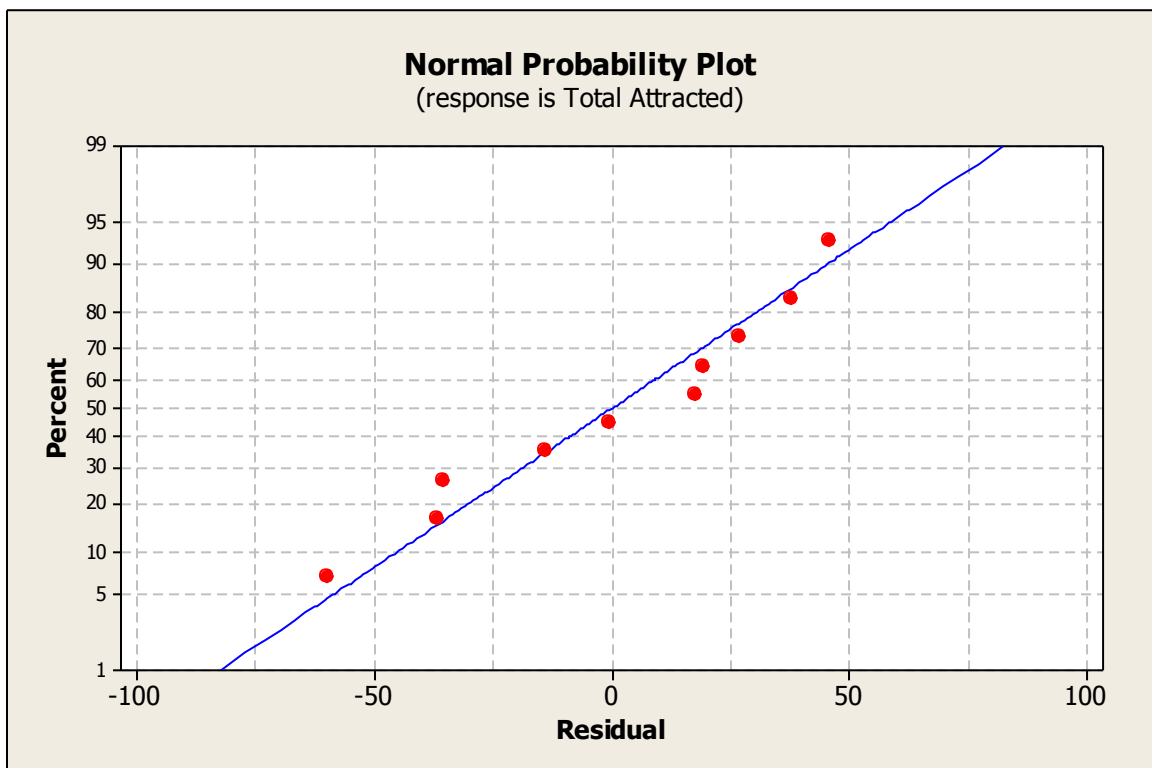
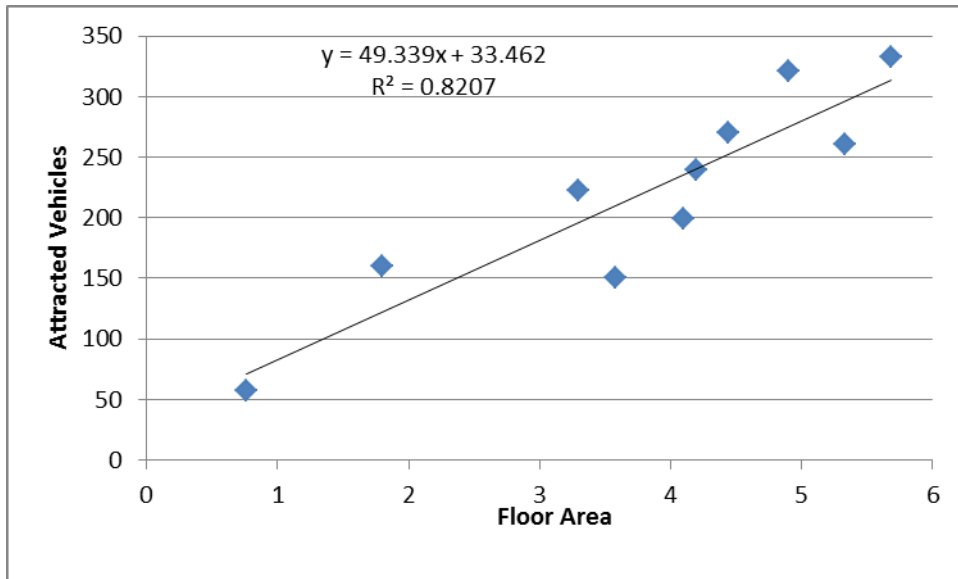
<i>Regression Statistics</i>	
Multiple R	0.96293
R Square	0.927235
Adjusted R Square	0.816124
Standard Error	66.8796
Observations	10

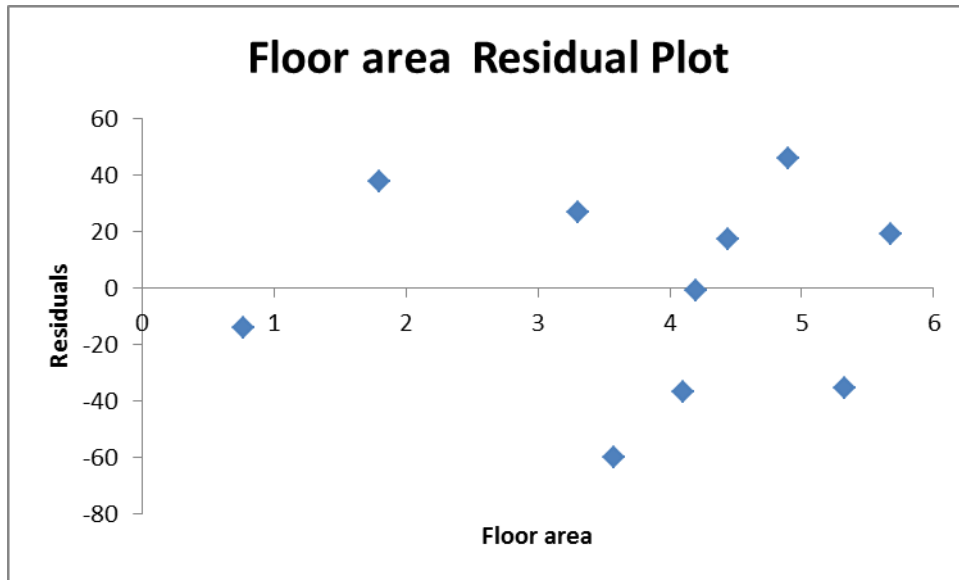
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	512974.1	512974.1	114.6854	5.07809E-06
Residual	9	40255.93	4472.881		
Total	10	553230			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Classes	15.5663675	1.453560871	10.70913	2.02E-06

Floor Area





SUMMARY OUTPUT

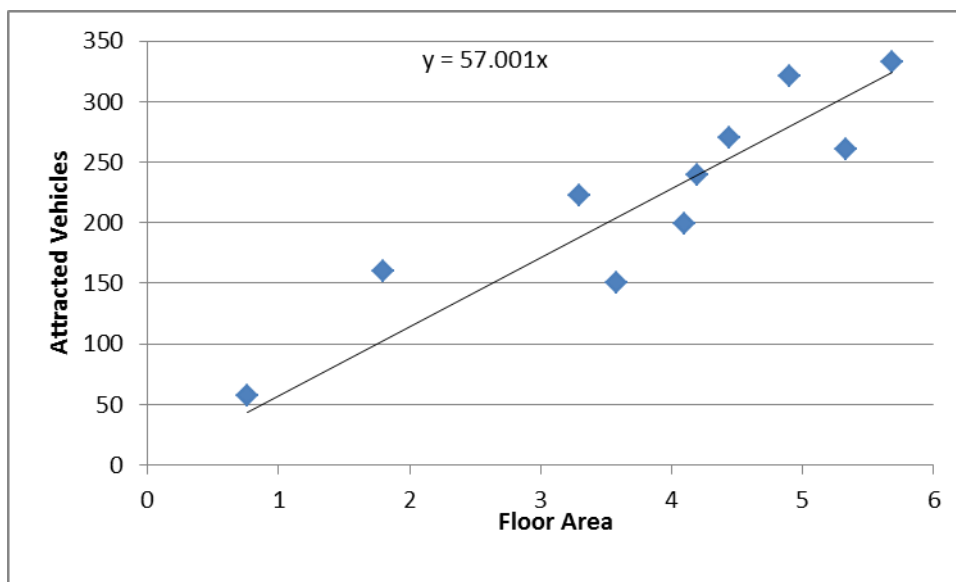
<i>Regression Statistics</i>	
Multiple R	0.905907
R Square	0.820667
Adjusted R Square	0.79825
Standard Error	37.59495
Observations	10

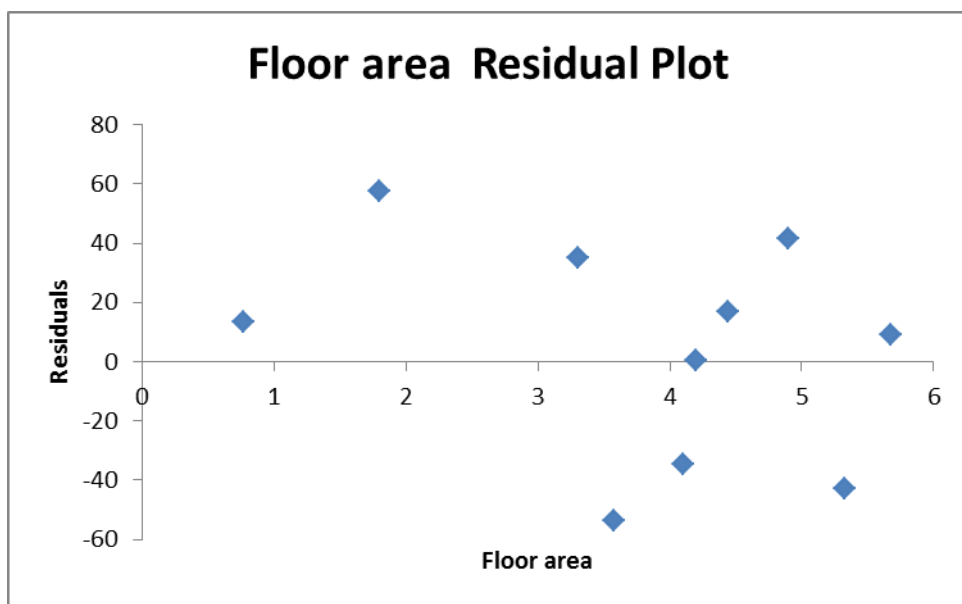
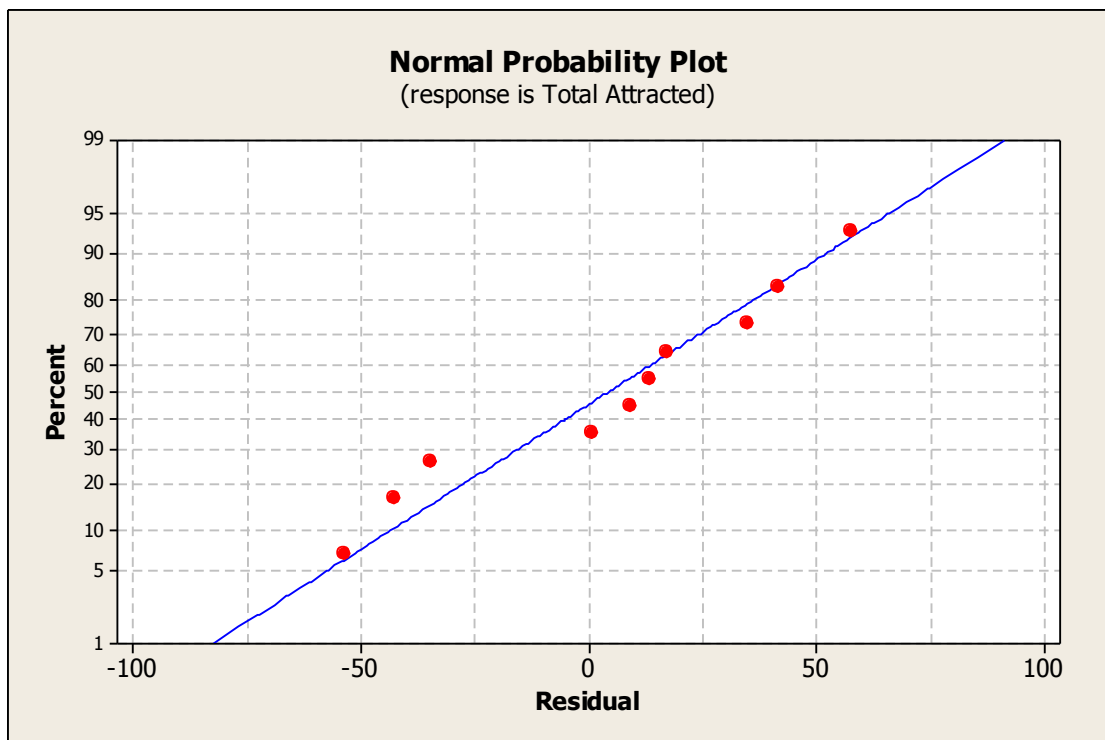
ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	51743.36	51743.36	36.60966	0.000305714
Residual	8	11307.04	1413.38		
Total	9	63050.4			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	33.46213007	33.25850735	1.006122	0.343815
Floor area	49.33917984	8.154439059	6.050592	0.000306

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below





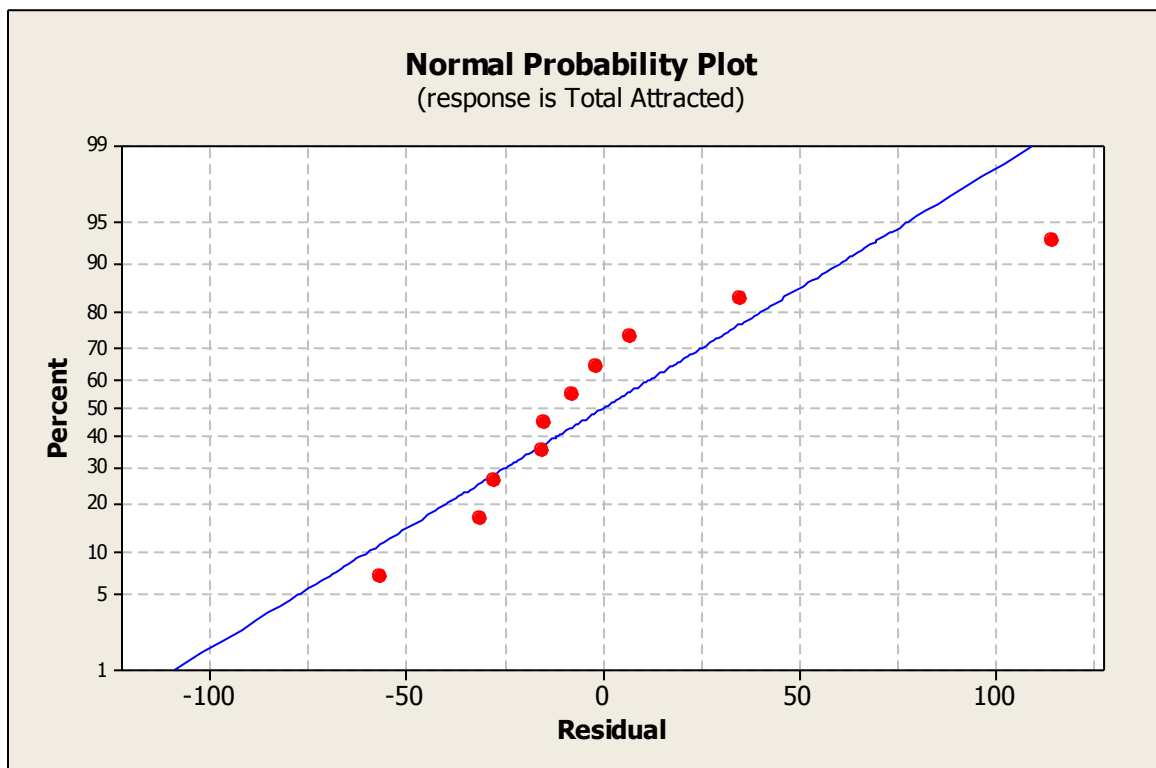
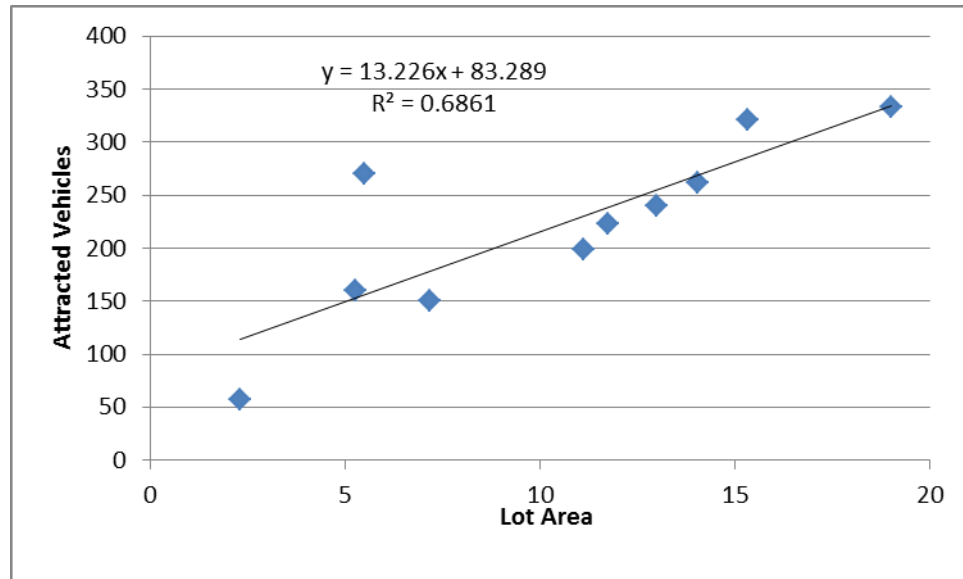
SUMMARY OUTPUT

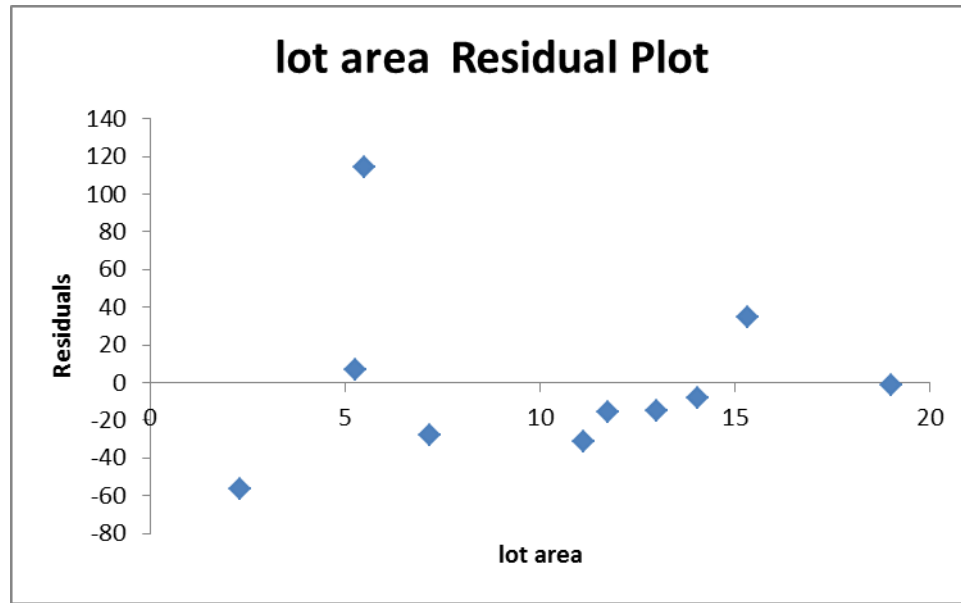
<i>Regression Statistics</i>	
Multiple R	0.988421
R Square	0.976976
Adjusted R Square	0.865865
Standard Error	37.62059
Observations	10

<i>ANOVA</i>					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	540492.2	540492.2	381.8899	4.88704E-08
Residual	9	12737.78	1415.309		
Total	10	553230			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Floor area	57.00147067	2.916869198	19.542	1.11E-08

Lot Area



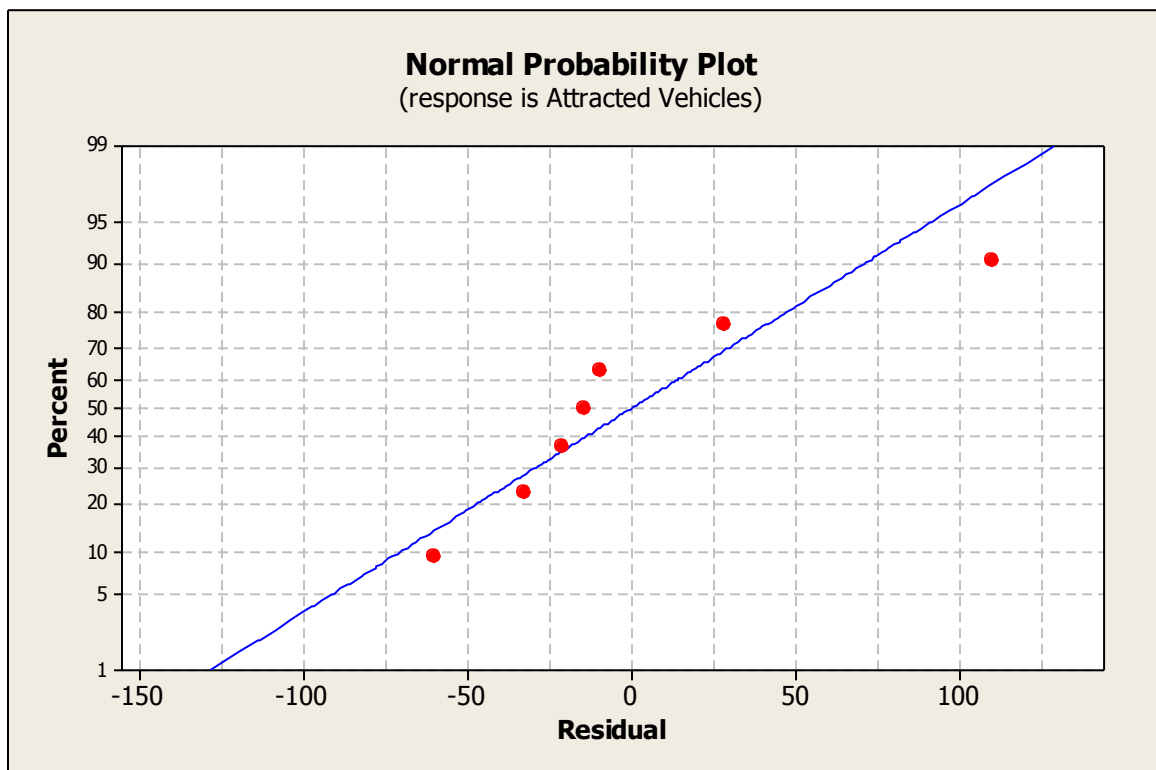
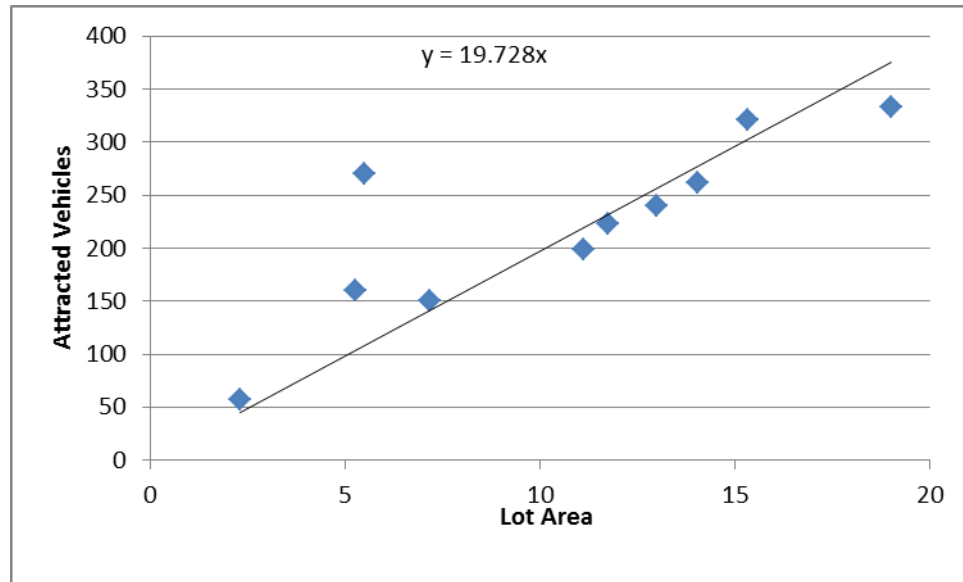


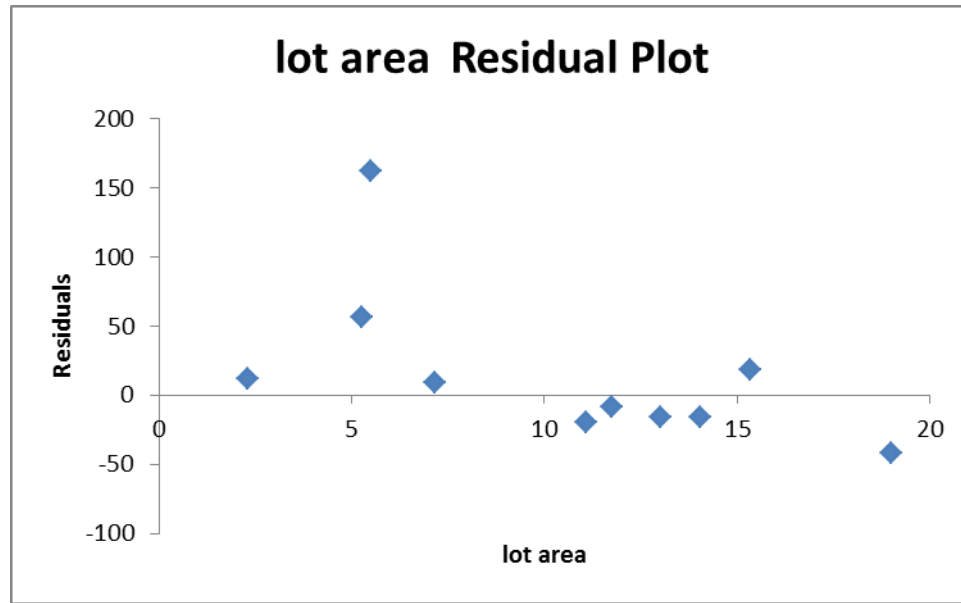
<i>Regression Statistics</i>	
Multiple R	0.828337
R Square	0.686141
Adjusted R Square	0.646909
Standard Error	49.73544
Observations	10

ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	43261.49	43261.49	17.48918	0.003071167
Residual	8	19788.91	2473.614		
Total	9	63050.4			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	83.28921881	36.57884453	2.276978	0.052318
Lot area	13.22646822	3.162708089	4.182007	0.003071





SUMMARY OUTPUT

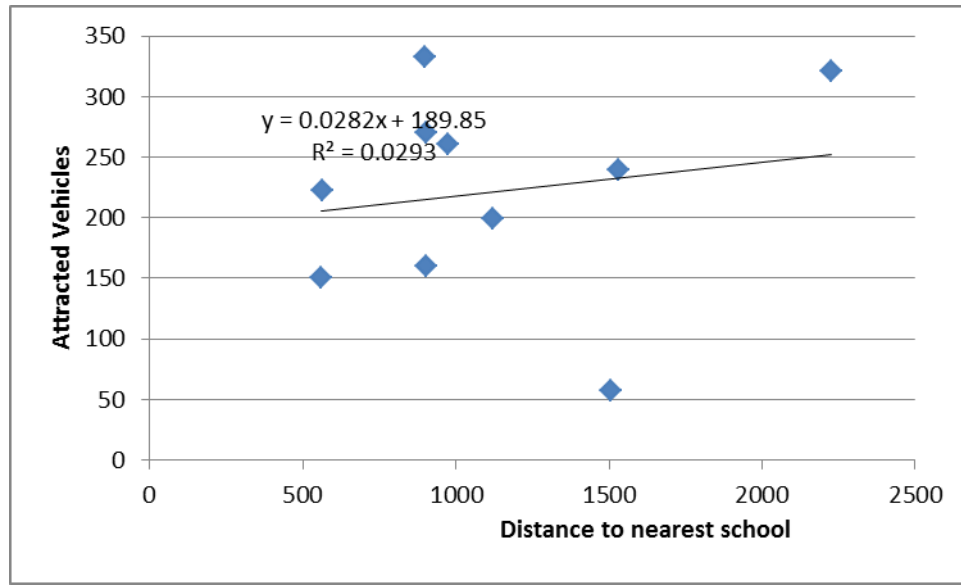
<i>Regression Statistics</i>	
Multiple R	0.970077
R Square	0.941049
Adjusted R Square	0.829937
Standard Error	60.19753
Observations	10

ANOVA

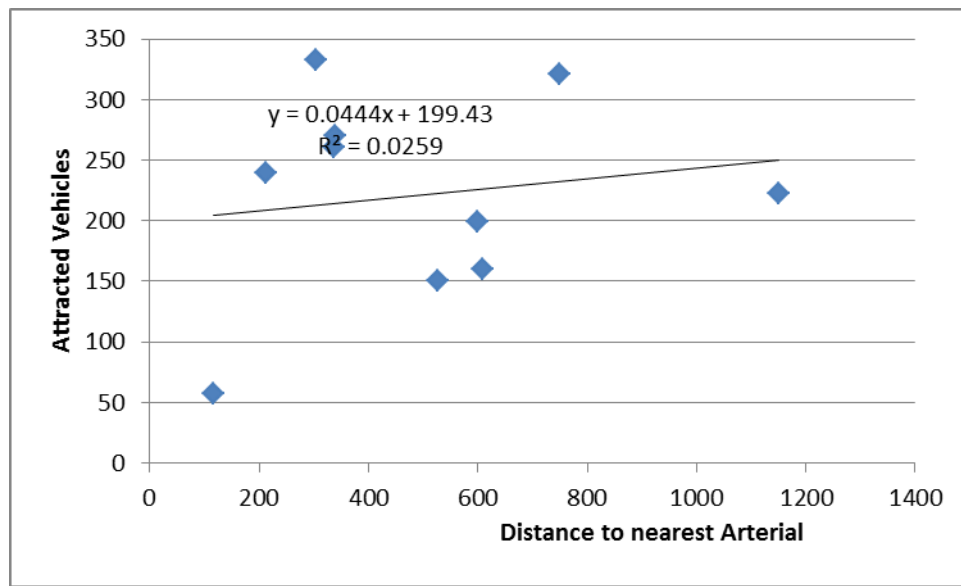
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	520616.3	520616.3	143.6682	2.1628E-06
Residual	9	32613.68	3623.742		
Total	10	553230			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Lot area	19.72822518	1.645916	11.98616	7.78E-07

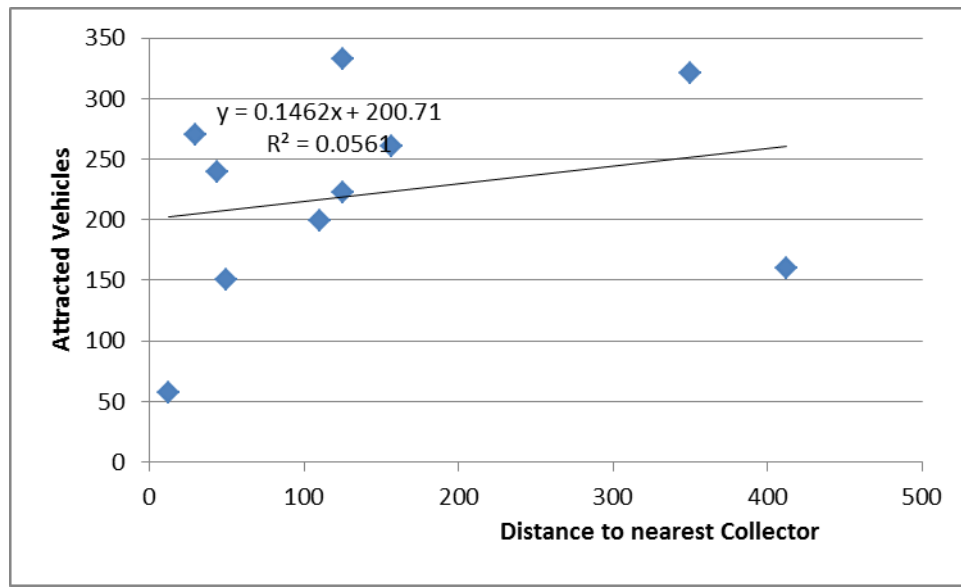
Distance to nearest school



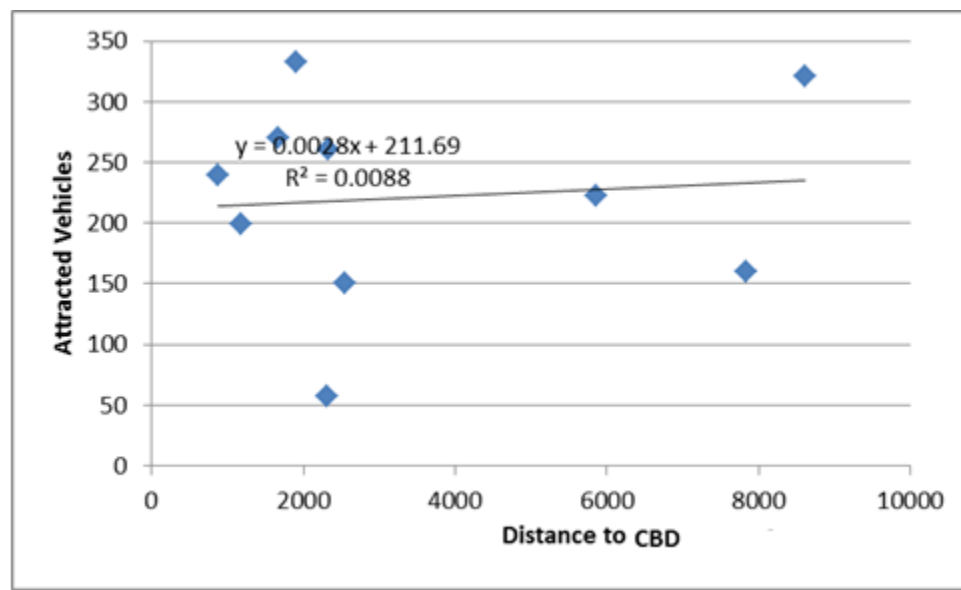
Distance to nearest arterial



Distance to nearest collector

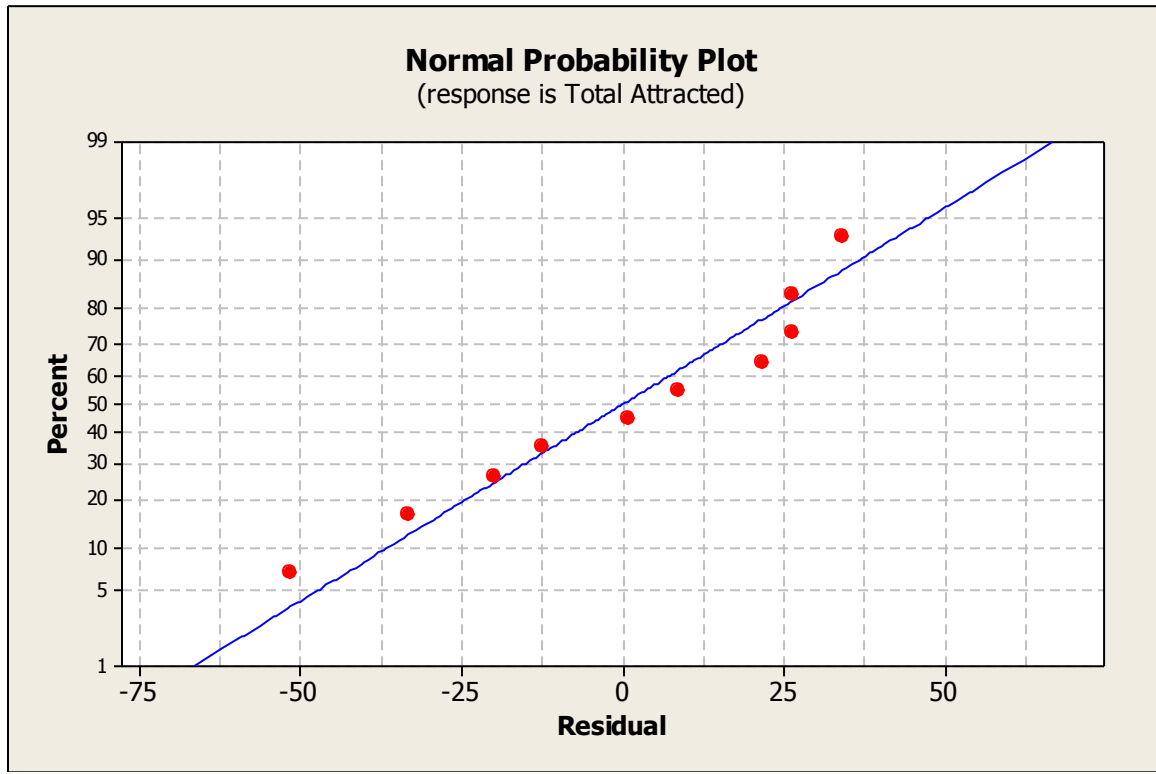


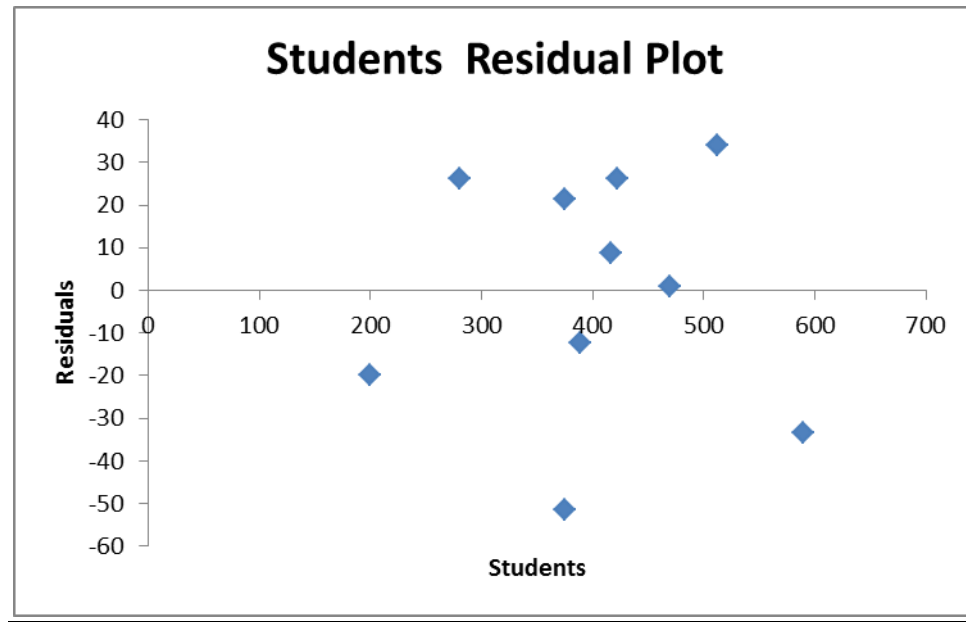
Distance to CBD



Multiple linear regressions using all of 10 schools

$$\text{Attracted vehicles} = -65.30 + 0.711 * \text{Students}$$





SUMMARY OUTPUT

<i>Regression Statistics</i>	
Multiple R	0.939635
R Square	0.882914
Adjusted R Square	0.868278
Standard Error	30.3775
Observations	10

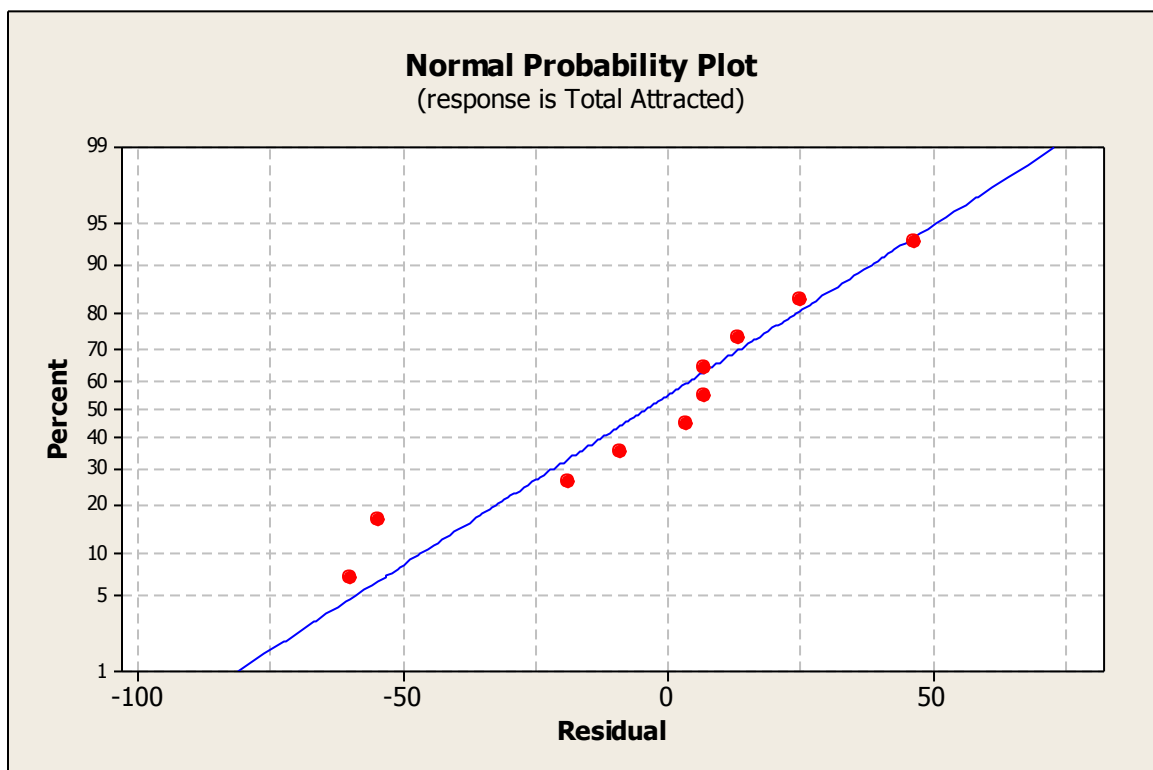
ANOVA

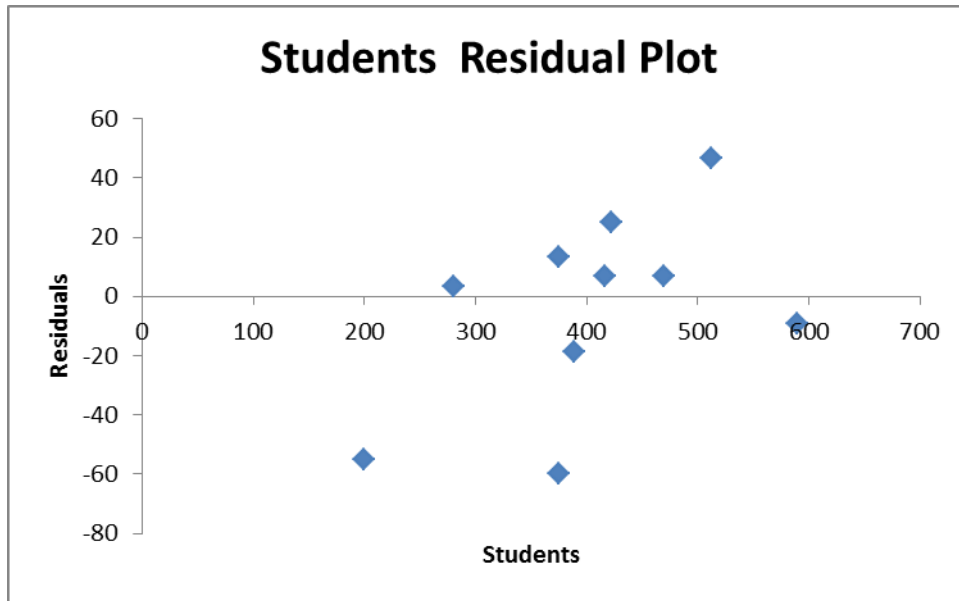
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	55668.06	55668.06	60.32567	5.39896E-05
Residual	8	7382.338	922.7922		
Total	9	63050.4			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	-65.292085	38.14126878	-1.71185	0.125284
Students	0.71139475	0.091592428	7.76696	5.4E-05

Since constant has a high P-value, the intercept was forced to be zero as shown in the second trial below

Attracted vehicles = 0.56*Students





SUMMARY OUTPUT

<i>Regression Statistics</i>	
Multiple R	0.990842
R Square	0.981768
Adjusted R Square	0.870657
Standard Error	33.47721
Observations	10

ANOVA

	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	1	543143.5	543143.5	484.6364	1.91405E-08
Residual	9	10086.51	1120.724		
Total	10	553230			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0	#N/A	#N/A	#N/A
Students	0.559656719	0.025422233	22.01446	3.89E-09

VITAE

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